

On Spectral Inclusion Sets and Computing the Spectra and Pseudospectra of Bounded Linear Operators

Simon Chandler-Wilde

University of Reading

1st July 2014 – Chicheley Hall

This talk is based on joint work with

- Ratchanikorn Chonchaiya, Reading
- Brian Davies, KCL
- **Marko Lindner**, TUHH, Germany

and supported by the Marie Curie Grants MEIF-CT-2005-009758 and PERG02-GA-2007-224761 of the European Union.

1 Introduction

- 1 Introduction
- 2 Inclusion Sets and Approximations for the Spectrum and Pseudospectrum

- 1 Introduction
- 2 Inclusion Sets and Approximations for the Spectrum and Pseudospectrum
- 3 Substantial example with lots of pictures! Feinberg-Zee Random Hopping Matrix

- 1 Introduction
- 2 Inclusion Sets and Approximations for the Spectrum and Pseudospectrum
- 3 Substantial example with lots of pictures! Feinberg-Zee Random Hopping Matrix
- 4 Extensions

- 1 Introduction
- 2 Inclusion Sets and Approximations for the Spectrum and Pseudospectrum
- 3 Substantial example with lots of pictures! Feinberg-Zee Random Hopping Matrix
- 4 Extensions

This talk is about a class of bounded linear operators on a space $E = \ell^p$ with $p \in [1, \infty]$.

So $x \in E$ iff $x = (x_k)_{k \in \mathbb{Z}}$, where $x_k \in \mathbb{C}$ for all $k \in \mathbb{Z}$ and

$$\begin{aligned}\|x\|_E &= \sqrt[p]{\sum_{k \in \mathbb{Z}} |x_k|^p}, & p < \infty, \\ \|x\|_E &= \sup_{k \in \mathbb{Z}} |x_k|, & p = \infty.\end{aligned}$$

$L(E) :=$ set of all **bounded** linear operators $E \rightarrow E$.

The Operators

$L(E) :=$ set of all **bounded** linear operators $E \rightarrow E$.

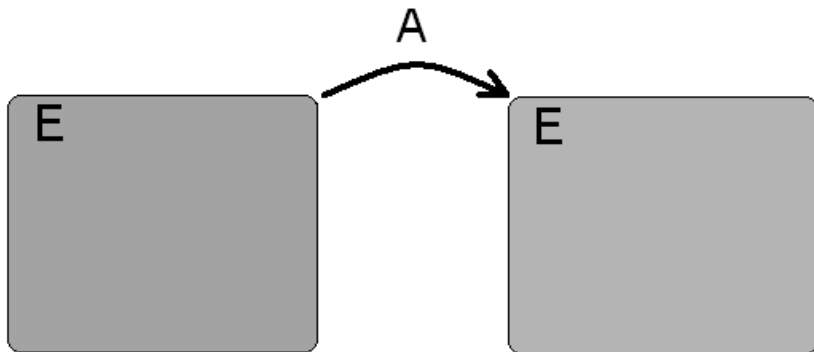
With every bounded linear operator A on E we will associate a matrix

$$\begin{pmatrix} \ddots & \vdots & \ddots \\ \cdots & a_{ij} & \cdots \\ \ddots & \vdots & \ddots \end{pmatrix} \begin{pmatrix} \vdots \\ x_j \\ \vdots \end{pmatrix} = \begin{pmatrix} \vdots \\ b_i \\ \vdots \end{pmatrix}$$

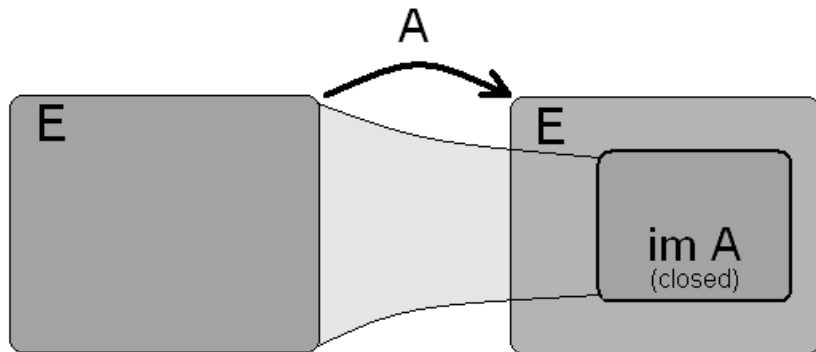
with indices $i, j \in \mathbb{Z}$ and entries $a_{ij} \in \mathbb{C}$.

For simplicity, we will restrict ourselves to **band matrices**.

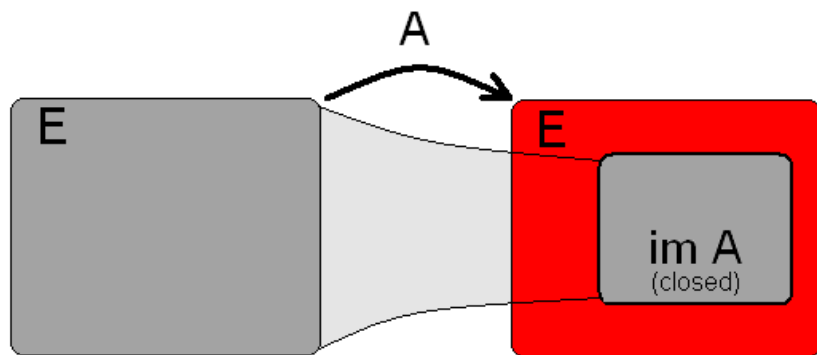
Fredholm operators



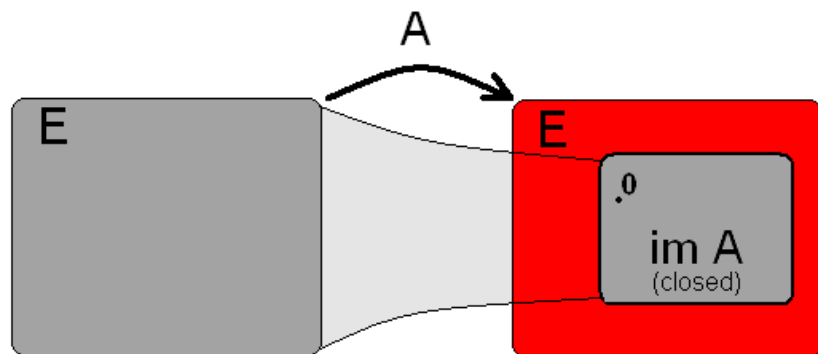
Fredholm operators



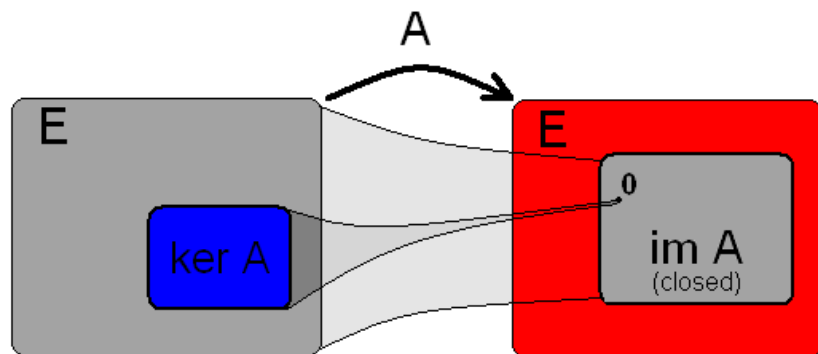
Fredholm operators

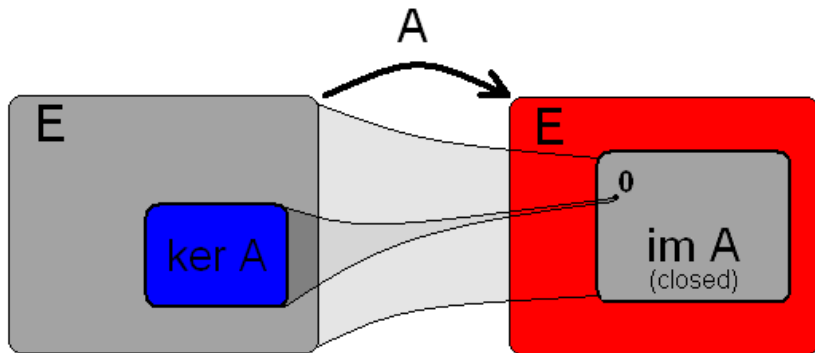


Fredholm operators



Fredholm operators





Definition

$A : E \rightarrow E$ is a *Fredholm operator*

if $\alpha := \dim(\ker A)$ and $\beta := \operatorname{codim}(\operatorname{im} A)$ are both finite.

The difference $\alpha - \beta$ is then called the *index of A* .

Spectrum, Essential Spectrum, Pseudospectra

For $A \in L(E)$ and $\varepsilon > 0$, we put

$$\operatorname{spec} A = \{\lambda \in \mathbb{C} : A - \lambda I \text{ not invertible on } E\},$$

Spectrum, Essential Spectrum, Pseudospectra

For $A \in L(E)$ and $\varepsilon > 0$, we put

$$\begin{aligned}\operatorname{spec} A &= \{\lambda \in \mathbb{C} : A - \lambda I \text{ not invertible on } E\}, \\ \operatorname{spec}_{\operatorname{ess}} A &= \{\lambda \in \mathbb{C} : A - \lambda I \text{ not Fredholm on } E\},\end{aligned}$$

Spectrum, Essential Spectrum, Pseudospectra

For $A \in L(E)$ and $\varepsilon > 0$, we put

$$\begin{aligned}\operatorname{spec} A &= \{\lambda \in \mathbb{C} : A - \lambda I \text{ not invertible on } E\}, \\ \operatorname{spec}_{\operatorname{ess}} A &= \{\lambda \in \mathbb{C} : A - \lambda I \text{ not Fredholm on } E\}, \\ \operatorname{spec}_{\varepsilon} A &= \operatorname{spec} A \cup \{\lambda \in \mathbb{C} : \|(A - \lambda I)^{-1}\| > 1/\varepsilon\}\end{aligned}$$

Spectrum, Essential Spectrum, Pseudospectra

For $A \in L(E)$ and $\varepsilon > 0$, we put

$$\begin{aligned}\operatorname{spec} A &= \{\lambda \in \mathbb{C} : A - \lambda I \text{ not invertible on } E\}, \\ \operatorname{spec}_{\operatorname{ess}} A &= \{\lambda \in \mathbb{C} : A - \lambda I \text{ not Fredholm on } E\}, \\ \operatorname{spec}_{\varepsilon} A &= \operatorname{spec} A \cup \{\lambda \in \mathbb{C} : \|(A - \lambda I)^{-1}\| > 1/\varepsilon\} \\ &= \bigcup_{\|T\| < \varepsilon} \operatorname{spec}(A + T)\end{aligned}$$

Spectrum, Essential Spectrum, Pseudospectra

For $A \in L(E)$ and $\varepsilon > 0$, we put

$$\begin{aligned}\operatorname{spec} A &= \{\lambda \in \mathbb{C} : A - \lambda I \text{ not invertible on } E\}, \\ \operatorname{spec}_{\operatorname{ess}} A &= \{\lambda \in \mathbb{C} : A - \lambda I \text{ not Fredholm on } E\}, \\ \operatorname{spec}_{\varepsilon} A &= \operatorname{spec} A \cup \{\lambda \in \mathbb{C} : \|(A - \lambda I)^{-1}\| > 1/\varepsilon\} \\ &= \bigcup_{\|T\| < \varepsilon} \operatorname{spec}(A + T) \\ &\supseteq \operatorname{spec} A + \varepsilon \mathbb{D} \quad (\text{"=" if } A \text{ is normal}).\end{aligned}$$

The sets $\operatorname{spec}_{\varepsilon} A$, $\varepsilon > 0$, are the so-called ε -**pseudospectra** of A . It holds that

$$\operatorname{spec} A =: \operatorname{spec}_0 A \subset \operatorname{spec}_{\varepsilon_1} A \subset \operatorname{spec}_{\varepsilon_2} A, \quad 0 < \varepsilon_1 < \varepsilon_2.$$

The Spectrum: Bounds and Approximations

- 1 Introduction
- 2 Inclusion Sets and Approximations for the Spectrum and Pseudospectrum
- 3 Substantial example with lots of pictures! Feinberg-Zee Random Hopping Matrix
- 4 Extensions

A Bi-Infinite Jacobi Matrix

We study bi-infinite matrices of the form

$$A = \begin{pmatrix} \ddots & & \ddots & & & & \\ & \ddots & \beta_{-2} & \gamma_{-1} & & & \\ & & \alpha_{-2} & \beta_{-1} & \gamma_0 & & \\ & & & \alpha_{-1} & \beta_0 & \gamma_1 & \\ & & & & \alpha_0 & \beta_1 & \gamma_1 \\ & & & & & \alpha_1 & \beta_2 & \ddots \\ & & & & & & \ddots & \ddots \end{pmatrix},$$

where $\alpha = (\alpha_i)$, $\beta = (\beta_i)$ and $\gamma = (\gamma_i)$ are bounded sequences of complex numbers.

A Bi-Infinite Jacobi Matrix

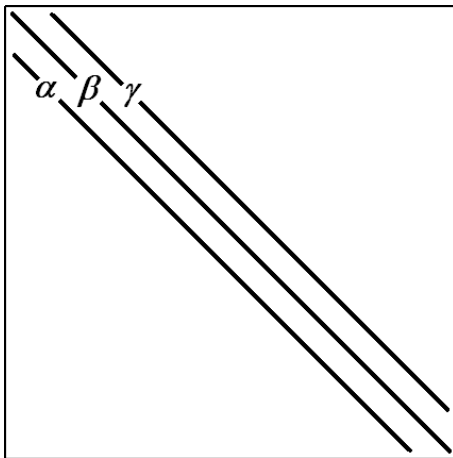
$$A = \begin{pmatrix} \ddots & \ddots & & & & \\ & \ddots & \beta_{-2} & \gamma_{-1} & & \\ & & \alpha_{-2} & \beta_{-1} & \gamma_0 & \\ & & & \alpha_{-1} & \beta_0 & \gamma_1 \\ & & & & \alpha_0 & \beta_1 & \gamma_1 \\ & & & & & \alpha_1 & \beta_2 & \ddots \\ & & & & & & \ddots & \ddots \end{pmatrix}$$

Task

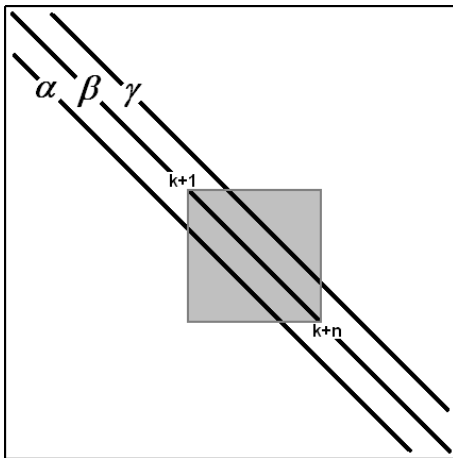
Compute **upper bounds on spectrum and pseudospectra** of A , understood as a bounded linear operator $\ell^2 \rightarrow \ell^2$.

Strategy

Look at (pseudo)spectra of the **finite principal submatrices** of A :

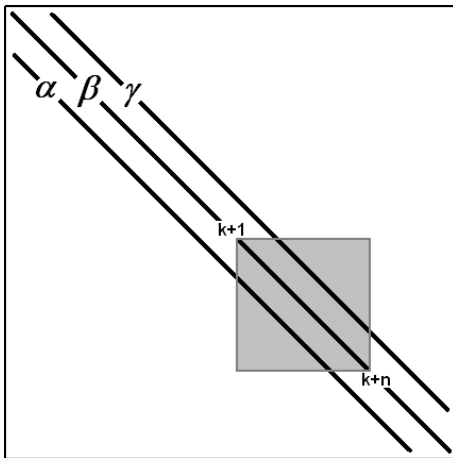


Look at (pseudo)spectra of the **finite principal submatrices** of A :



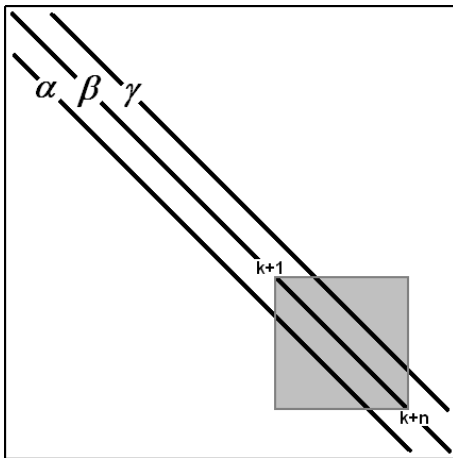
Strategy

Look at (pseudo)spectra of the **finite principal submatrices** of A :



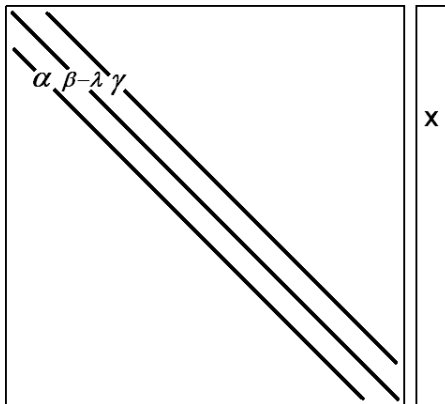
Strategy

Look at (pseudo)spectra of the **finite principal submatrices** of A :



Method 1: Finite principal submatrices

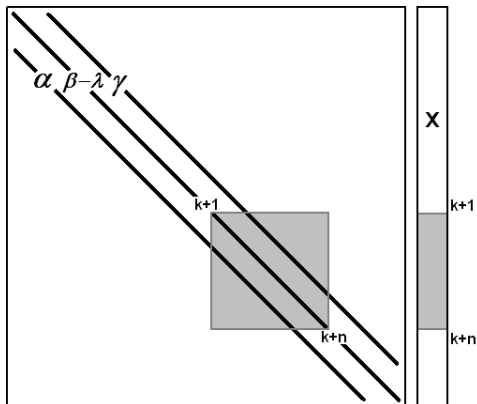
Let $\lambda \in \text{spec}_\varepsilon(A)$ and let $x \in \ell^2$ be a corresponding pseudomode.



$$\|(A - \lambda I)x\| < \varepsilon \|x\|$$

Method 1: Finite principal submatrices

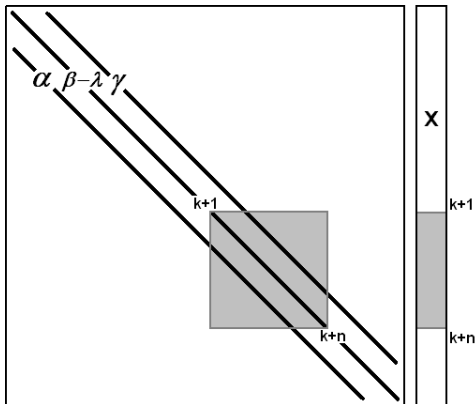
Let $\lambda \in \text{spec}_\varepsilon(A)$ and let $x \in \ell^2$ be a corresponding pseudomode.



$$\|(A - \lambda I)x\| < \varepsilon \|x\|$$

Method 1: Finite principal submatrices

Let $\lambda \in \text{spec}_\varepsilon(A)$ and let $x \in \ell^2$ be a corresponding pseudomode.



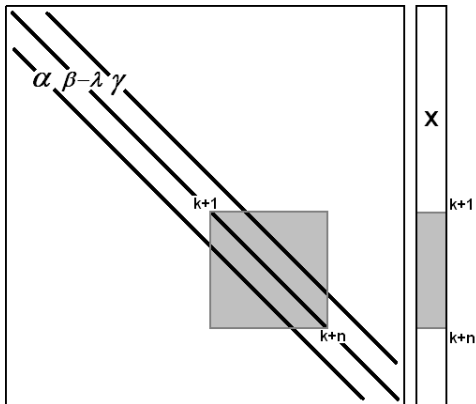
$$\|(A - \lambda I)x\| < \varepsilon \|x\|$$

Claim: $\exists k \in \mathbb{Z} :$

$$\begin{aligned} \|(A_{n,k} - \lambda I_n)x_{n,k}\| \\ < (\varepsilon + \varepsilon_n) \|x_{n,k}\| \end{aligned}$$

Method 1: Finite principal submatrices

Let $\lambda \in \text{spec}_\varepsilon(A)$ and let $x \in \ell^2$ be a corresponding pseudomode.



$$\|(A - \lambda I)x\| < \varepsilon \|x\|$$

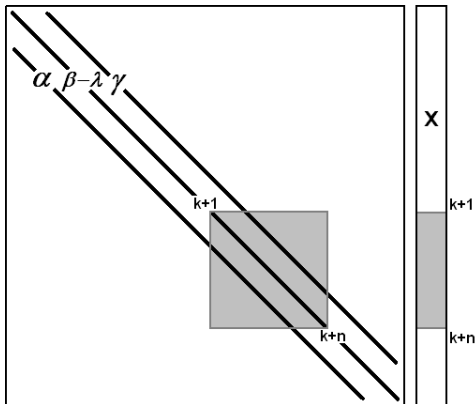
Claim: $\exists k \in \mathbb{Z}$:

$$\begin{aligned} \|(A_{n,k} - \lambda I_n)x_{n,k}\| \\ < (\varepsilon + \varepsilon_n) \|x_{n,k}\| \end{aligned}$$

$$\begin{aligned} \sum_k \|(A_{n,k} - \lambda I_n)x_{n,k}\|^2 \\ < (\varepsilon + \varepsilon_n)^2 \sum_k \|x_{n,k}\|^2 \end{aligned}$$

Method 1: Finite principal submatrices

Let $\lambda \in \text{spec}_\varepsilon(A)$ and let $x \in \ell^2$ be a corresponding pseudomode.



$$\|(A - \lambda I)x\| < \varepsilon \|x\|$$

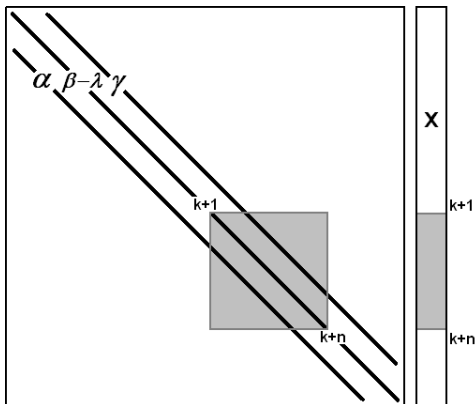
Claim: $\exists k \in \mathbb{Z} :$

$$\begin{aligned} \|(A_{n,k} - \lambda I_n)x_{n,k}\| \\ < (\varepsilon + \varepsilon_n) \|x_{n,k}\| \end{aligned}$$

$$\begin{aligned} \|(A_{n,k} - \lambda I_n)x_{n,k}\|^2 \\ < (\varepsilon + \varepsilon_n)^2 \|x_{n,k}\|^2 \end{aligned}$$

Method 1: Finite principal submatrices

Let $\lambda \in \text{spec}_\varepsilon(A)$ and let $x \in \ell^2$ be a corresponding pseudomode.

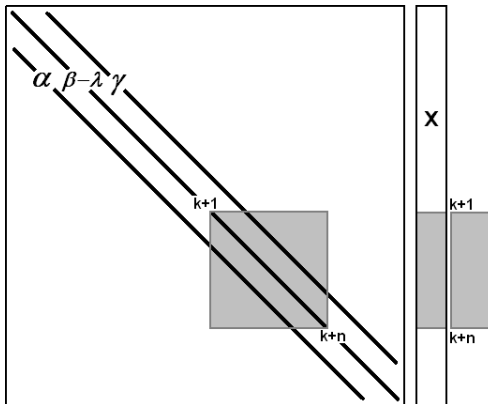


$$\|(A - \lambda I)x\| < \varepsilon \|x\|$$

$$\begin{aligned} \|(A_{n,k} - \lambda I_n)x_{n,k}\| \\ < (\varepsilon + \varepsilon_n) \|x_{n,k}\| \end{aligned}$$

Method 1: Finite principal submatrices

Let $\lambda \in \text{spec}_\varepsilon(A)$ and let $x \in \ell^2$ be a corresponding pseudomode.



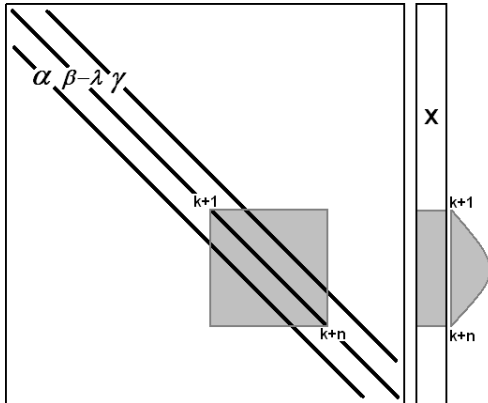
$$\|(A - \lambda I)x\| < \varepsilon \|x\|$$

$$\begin{aligned} \|(A_{n,k} - \lambda I_n)x_{n,k}\| \\ < (\varepsilon + \varepsilon_n) \|x_{n,k}\| \end{aligned}$$

$$\varepsilon_n < \frac{1}{\sqrt{n}} (\|\alpha\|_\infty + \|\gamma\|_\infty)$$

Method 1: Finite principal submatrices

Let $\lambda \in \text{spec}_\varepsilon(A)$ and let $x \in \ell^2$ be a corresponding pseudomode.



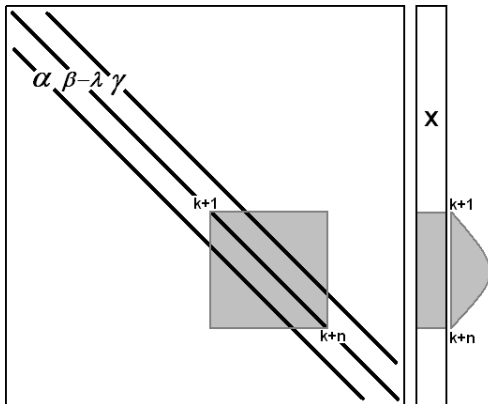
$$\|(A - \lambda I)x\| < \varepsilon \|x\|$$

$$\begin{aligned} \|(A_{n,k} - \lambda I_n)x_{n,k}\| \\ < (\varepsilon + \varepsilon_n) \|x_{n,k}\| \end{aligned}$$

$$\varepsilon_n < \frac{\pi}{n} (\|\alpha\|_\infty + \|\gamma\|_\infty)$$

Method 1: Finite principal submatrices

Let $\lambda \in \text{spec}_\varepsilon(A)$ and let $x \in \ell^2$ be a corresponding pseudomode.



$$\|(A - \lambda I)x\| < \varepsilon \|x\|$$

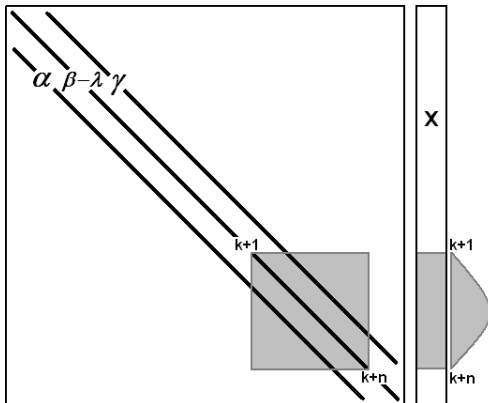
$$\begin{aligned} \|(A_{n,k} - \lambda I_n)x_{n,k}\| \\ < (\varepsilon + \varepsilon_n) \|x_{n,k}\| \end{aligned}$$

$$\varepsilon_n < \frac{\pi}{n} (\|\alpha\|_\infty + \|\gamma\|_\infty)$$

$$\Rightarrow \lambda \in \text{spec}_{\varepsilon + \varepsilon_n}(A_{n,k})$$

Method 1: Finite principal submatrices

Let $\lambda \in \text{spec}_\varepsilon(A)$ and let $x \in \ell^2$ be a corresponding pseudomode.



$$\|(A - \lambda I)x\| < \varepsilon \|x\|$$

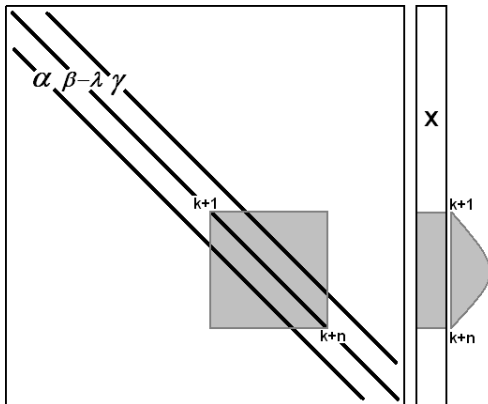
$$\begin{aligned} \|(A_{n,k} - \lambda I_n)x_{n,k}\| \\ < (\varepsilon + \varepsilon_n) \|x_{n,k}\| \end{aligned}$$

$$\varepsilon_n < \frac{\pi}{n} (\|\alpha\|_\infty + \|\gamma\|_\infty)$$

$$\Rightarrow \lambda \in \text{spec}_{\varepsilon + \varepsilon_n}(A_{n,k})$$

Method 1: Finite principal submatrices

Let $\lambda \in \text{spec}_\varepsilon(A)$ and let $x \in \ell^2$ be a corresponding pseudomode.



$$\|(A - \lambda I)x\| < \varepsilon \|x\|$$

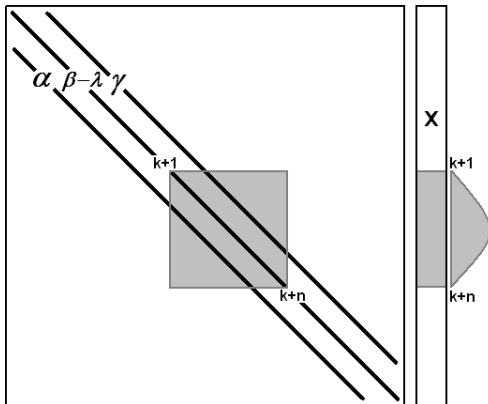
$$\begin{aligned} \|(A_{n,k} - \lambda I_n)x_{n,k}\| \\ < (\varepsilon + \varepsilon_n) \|x_{n,k}\| \end{aligned}$$

$$\varepsilon_n < \frac{\pi}{n} (\|\alpha\|_\infty + \|\gamma\|_\infty)$$

$$\Rightarrow \lambda \in \text{spec}_{\varepsilon + \varepsilon_n}(A_{n,k})$$

Method 1: Finite principal submatrices

Let $\lambda \in \text{spec}_\varepsilon(A)$ and let $x \in \ell^2$ be a corresponding pseudomode.



$$\|(A - \lambda I)x\| < \varepsilon \|x\|$$

$$\begin{aligned} \|(A_{n,k} - \lambda I_n)x_{n,k}\| \\ < (\varepsilon + \varepsilon_n) \|x_{n,k}\| \end{aligned}$$

$$\varepsilon_n < \frac{\pi}{n} (\|\alpha\|_\infty + \|\gamma\|_\infty)$$

$$\Rightarrow \lambda \in \text{spec}_{\varepsilon + \varepsilon_n}(A_{n,k})$$

Method 1: Finite principal submatrices

So one gets

Upper Bound

$$\operatorname{spec}_{\varepsilon}(A) \subset \bigcup_{k \in \mathbb{Z}} \operatorname{spec}_{\varepsilon + \varepsilon_n}(A_{n,k}),$$

where

$$\varepsilon_n = 2(\|\alpha\|_{\infty} + \|\gamma\|_{\infty}) \sin \frac{\theta}{2} < \frac{\pi}{n+2}(\|\alpha\|_{\infty} + \|\gamma\|_{\infty}),$$

and $\frac{\pi}{2n+1} \leq \theta \leq \frac{\pi}{n+2}$ satisfies

$$2 \sin \frac{\theta}{2} \cos(n + \frac{1}{2})\theta + \frac{\|\alpha\|_{\infty} \|\gamma\|_{\infty}}{(\|\alpha\|_{\infty} + \|\gamma\|_{\infty})^2} \sin(n-1)\theta = 0.$$

Method 1: Finite principal submatrices

So one gets

Upper Bound

$$\operatorname{spec}_{\varepsilon}(A) \subset \bigcup_{k \in \mathbb{Z}} \operatorname{spec}_{\varepsilon + \varepsilon_n}(A_{n,k}),$$

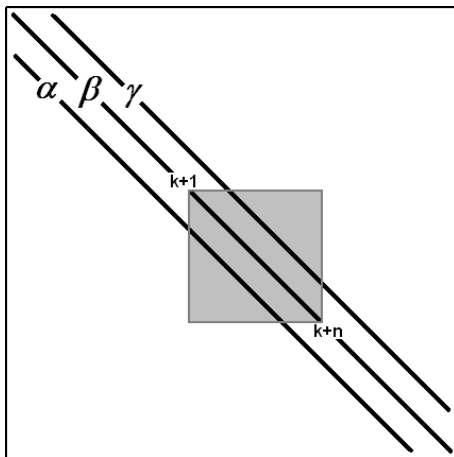
where

$$\varepsilon_n = 2(\|\alpha\|_{\infty} + \|\gamma\|_{\infty}) \sin \frac{\pi}{4n+2} < \frac{\pi}{2n+1}(\|\alpha\|_{\infty} + \|\gamma\|_{\infty}),$$

if $\alpha = 0$ or $\gamma = 0$, i.e., A is **bi-diagonal**.

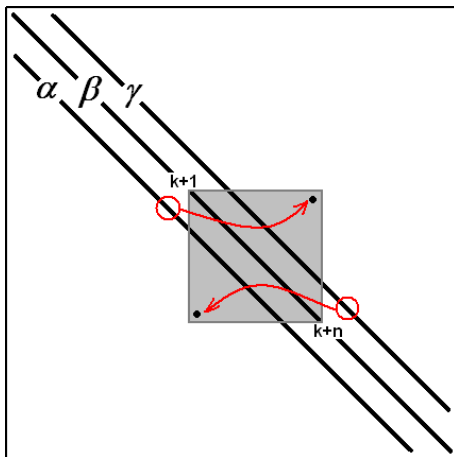
Method 2: **Periodised** finite principal submatrices

If the finite submatrices $A_{n,k}$ are “periodised”,



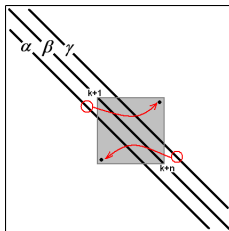
Method 2: **Periodised** finite principal submatrices

If the finite submatrices $A_{n,k}$ are “periodised”,



Method 2: **Periodised** finite principal submatrices

If the finite submatrices $A_{n,k}$ are “periodised”,



very similar computations show that, again,

$$\text{spec}_{\varepsilon}(A) \subset \bigcup_{k \in \mathbb{Z}} \text{spec}_{\varepsilon + \varepsilon_n}(A_{n,k}^{\text{per}})$$

$$\text{with } \varepsilon_n = 2(\|\alpha\|_{\infty} + \|\gamma\|_{\infty}) \sin \frac{\pi}{2n} < \frac{\pi}{n}(\|\alpha\|_{\infty} + \|\gamma\|_{\infty})$$

and this upper bound on $\text{spec}_{\varepsilon}(A)$ can be sharper than method 1.

Method 1 vs. Method 2: An Example

Look at the shift operator

$$(Ax)(i) = x(i+1), \quad \text{i.e.} \quad A = \begin{pmatrix} \cdot & \cdot & & & \\ & 0 & 1 & & \\ & & 0 & 1 & \\ & & & 0 & \ddots \\ & & & & \ddots \end{pmatrix}.$$

Method 1 vs. Method 2: An Example

Look at the shift operator

$$(Ax)(i) = x(i+1), \quad \text{i.e.} \quad A = \begin{pmatrix} & & & & \\ & & & & \\ & & 0 & 1 & \\ & & & 0 & 1 \\ & & & & 0 & \ddots \end{pmatrix}.$$

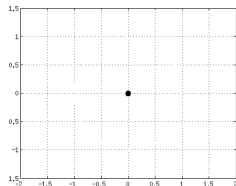
$$A_{n,k} = \begin{pmatrix} & & & & & \\ & & & & & \\ & & 0 & 1 & & \\ & & & 0 & 1 & \\ & & & & 0 & 1 \\ & & & & & 0 \end{pmatrix}$$

Method 1 vs. Method 2: An Example

Look at the shift operator

$$(Ax)(i) = x(i+1), \quad \text{i.e.} \quad A = \begin{pmatrix} & & & & \\ & & & & \\ & 0 & 1 & & \\ & & 0 & 1 & \\ & & & 0 & \ddots \end{pmatrix}.$$

$$A_{n,k} = \begin{pmatrix} & 0 & & & & \\ & & 1 & & & \\ & & & 1 & & \\ & 0 & & 0 & 1 & \\ & & & 0 & & 1 \\ & & & & 0 & \end{pmatrix} \quad \text{spec } A_{n,k} =$$

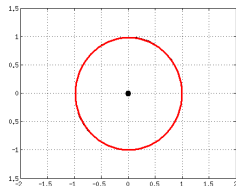


Method 1 vs. Method 2: An Example

Look at the shift operator

$$(Ax)(i) = x(i+1), \quad \text{i.e.} \quad A = \begin{pmatrix} & & & & \\ & & & & \\ & & 0 & 1 & \\ & & & 0 & 1 \\ & & & & 0 & \ddots \end{pmatrix}.$$

$$A_{n,k} = \begin{pmatrix} & & & & \\ & & & & \\ & & 0 & 1 & \\ & & & 0 & 1 \\ & & & & 0 & \ddots \end{pmatrix} \quad \text{spec } A_{n,k} =$$



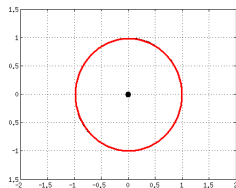
Method 1 vs. Method 2: An Example

Look at the shift operator

$$(Ax)(i) = x(i+1), \quad \text{i.e.} \quad A = \begin{pmatrix} & & & & \\ & & & & \\ & & 0 & 1 & \\ & & & 0 & 1 \\ & & & & 0 \\ & & & & & \ddots \end{pmatrix}.$$

$$A_{n,k} = \begin{pmatrix} 0 & 1 & & & \\ & 0 & 1 & & \\ & & 0 & 1 & \\ & & & 0 & 1 \\ & & & & 0 \end{pmatrix} \quad \text{spec } A_{n,k} =$$

$$A_{n,k}^{\text{per}} = \begin{pmatrix} 0 & 1 & & & \\ & 0 & 1 & & \\ & & 0 & 1 & \\ & & & 0 & 1 \\ 1 & & & & 0 \end{pmatrix}$$



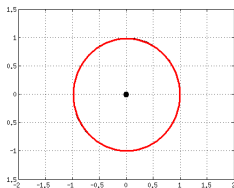
Method 1 vs. Method 2: An Example

Look at the shift operator

$$(Ax)(i) = x(i+1), \quad \text{i.e.} \quad A = \begin{pmatrix} & & & & \\ & & & & \\ & & 0 & 1 & \\ & & & 0 & 1 \\ & & & & 0 \\ & & & & & \ddots \end{pmatrix}.$$

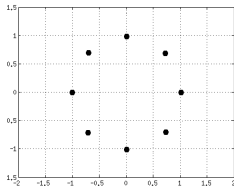
$$A_{n,k} = \begin{pmatrix} & & & & \\ & & & & \\ & & 0 & 1 & \\ & & & 0 & 1 \\ & & & & 0 \\ & & & & & \ddots \end{pmatrix}$$

$$\text{spec } A_{n,k} =$$



$$A_{n,k}^{\text{per}} = \begin{pmatrix} & & & & \\ & & & & \\ & & 0 & 1 & \\ & & & 0 & 1 \\ & & & & 0 \\ & & & & & \ddots \\ 1 & & & & & \end{pmatrix}$$

$$\text{spec } A_{n,k}^{\text{per}} =$$



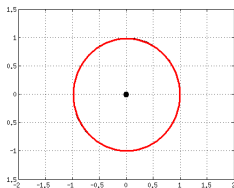
Method 1 vs. Method 2: An Example

Look at the shift operator

$$(Ax)(i) = x(i+1), \quad \text{i.e.} \quad A = \begin{pmatrix} & & & & \\ & & & & \\ & & 0 & 1 & \\ & & & 0 & 1 \\ & & & & 0 \\ & & & & & \ddots \end{pmatrix}.$$

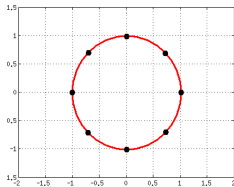
$$A_{n,k} = \begin{pmatrix} & & & & \\ & & & & \\ & & 0 & 1 & \\ & & & 0 & 1 \\ & & & & 0 \\ & & & & & \ddots \end{pmatrix}$$

$$\text{spec } A_{n,k} =$$



$$A_{n,k}^{\text{per}} = \begin{pmatrix} & & & & \\ & & & & \\ & & 0 & 1 & \\ & & & 0 & 1 \\ & & & & 0 \\ & & & & & \ddots \\ 1 & & & & & \end{pmatrix}$$

$$\text{spec } A_{n,k}^{\text{per}} =$$

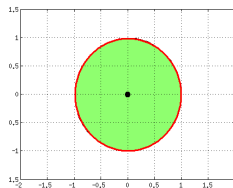


Method 1 vs. Method 2: An Example

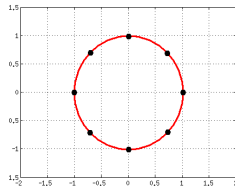
Look at the shift operator

$$(Ax)(i) = x(i+1), \quad \text{i.e.} \quad A = \begin{pmatrix} 0 & 1 & & \\ & 0 & 1 & \\ & & 0 & \ddots \\ & & & 0 \end{pmatrix}.$$

$$A_{n,k} = \begin{pmatrix} 0 & 1 & & & \\ & 0 & 1 & & \\ & & 0 & 1 & \\ & & & 0 & 1 \\ 1 & & & & 0 \end{pmatrix} \quad \text{spec}_{\varepsilon_n} A_{n,k} =$$



$$A_{n,k}^{\text{per}} = \begin{pmatrix} 0 & 1 & & & \\ & 0 & 1 & & \\ & & 0 & 1 & \\ & & & 0 & 1 \\ 1 & & & & 0 \end{pmatrix} \quad \text{spec } A_{n,k}^{\text{per}} =$$



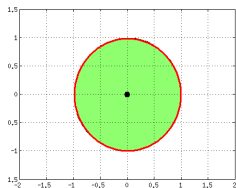
Method 1 vs. Method 2: An Example

Look at the shift operator

$$(Ax)(i) = x(i+1), \quad \text{i.e.} \quad A = \begin{pmatrix} 0 & 1 & & \\ & 0 & 1 & \\ & & 0 & \ddots \\ & & & 0 \end{pmatrix}.$$

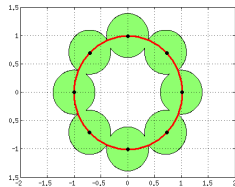
$$A_{n,k} = \begin{pmatrix} 0 & 1 & & & \\ & 0 & 1 & & \\ & & 0 & 1 & \\ & & & 0 & 1 \\ 1 & & & & 0 \end{pmatrix}$$

$$\text{spec}_{\varepsilon_n} A_{n,k} =$$



$$A_{n,k}^{\text{per}} = \begin{pmatrix} 0 & 1 & & & \\ & 0 & 1 & & \\ & & 0 & 1 & \\ & & & 0 & 1 \\ 1 & & & & 0 \end{pmatrix}$$

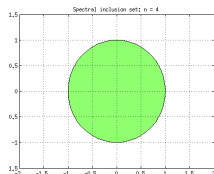
$$\text{spec}_{\varepsilon_n} A_{n,k}^{\text{per}} =$$



Method 1 vs. Method 2: An Example

Look at the shift operator

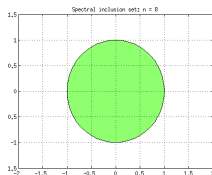
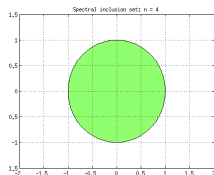
$$(Ax)(i) = x(i+1), \quad \text{i.e.} \quad A = \begin{pmatrix} \cdot & \cdot & & \\ & 0 & 1 & \\ & & 0 & 1 & \\ & & & 0 & \ddots \end{pmatrix}.$$



Method 1 vs. Method 2: An Example

Look at the shift operator

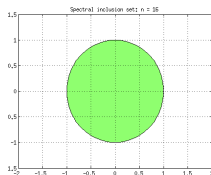
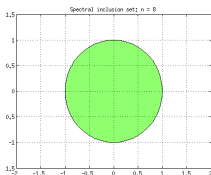
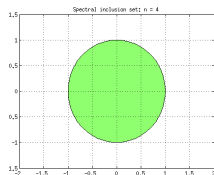
$$(Ax)(i) = x(i+1), \quad \text{i.e.} \quad A = \begin{pmatrix} \cdot & \cdot & & \\ & 0 & 1 & \\ & & 0 & 1 & \\ & & & 0 & \cdot \end{pmatrix}.$$



Method 1 vs. Method 2: An Example

Look at the shift operator

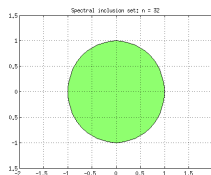
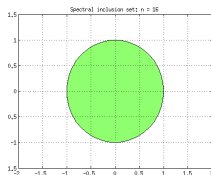
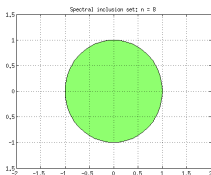
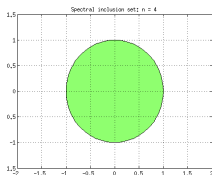
$$(Ax)(i) = x(i+1), \quad \text{i.e.} \quad A = \begin{pmatrix} & & & \\ & 0 & 1 & \\ & & 0 & 1 \\ & & & 0 & \ddots \end{pmatrix}.$$



Method 1 vs. Method 2: An Example

Look at the shift operator

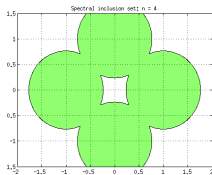
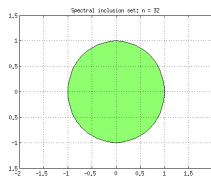
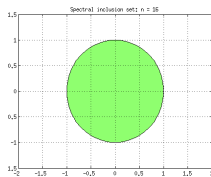
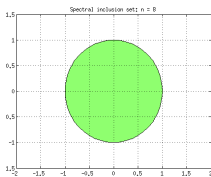
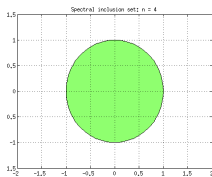
$$(Ax)(i) = x(i+1), \quad \text{i.e.} \quad A = \begin{pmatrix} & & & \\ & 0 & 1 & \\ & & 0 & 1 \\ & & & 0 \end{pmatrix}.$$



Method 1 vs. Method 2: An Example

Look at the shift operator

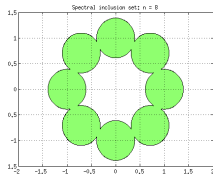
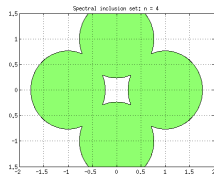
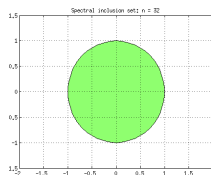
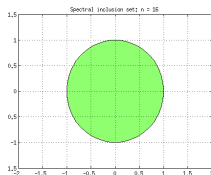
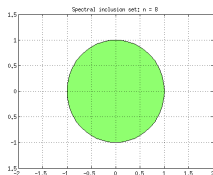
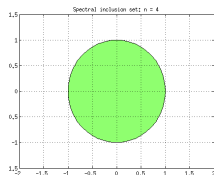
$$(Ax)(i) = x(i+1), \quad \text{i.e.} \quad A = \begin{pmatrix} & & & \\ & 0 & 1 & \\ & & 0 & 1 \\ & & & 0 \end{pmatrix}.$$



Method 1 vs. Method 2: An Example

Look at the shift operator

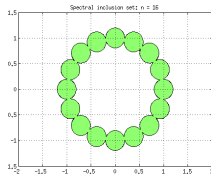
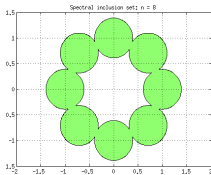
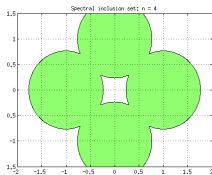
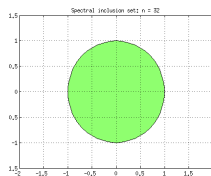
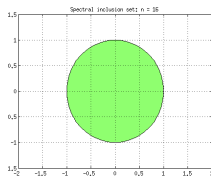
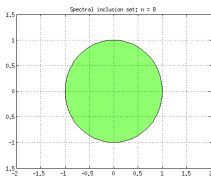
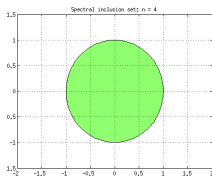
$$(Ax)(i) = x(i+1), \quad \text{i.e.} \quad A = \begin{pmatrix} & & & \\ & 0 & 1 & \\ & & 0 & 1 \\ & & & 0 \end{pmatrix}.$$



Method 1 vs. Method 2: An Example

Look at the shift operator

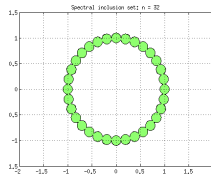
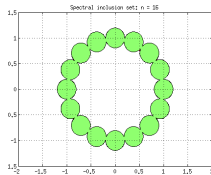
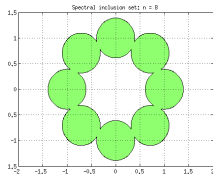
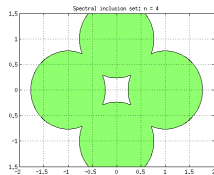
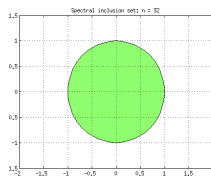
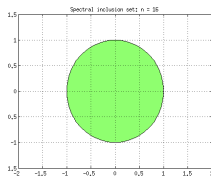
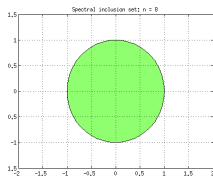
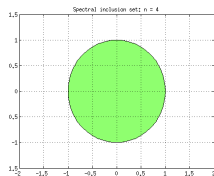
$$(Ax)(i) = x(i+1), \quad \text{i.e.} \quad A = \begin{pmatrix} & & & \\ & 0 & 1 & \\ & & 0 & 1 \\ & & & 0 \end{pmatrix}.$$



Method 1 vs. Method 2: An Example

Look at the shift operator

$$(Ax)(i) = x(i+1), \quad \text{i.e.} \quad A = \begin{pmatrix} & & & & \\ & & & & \\ & & 0 & 1 & \\ & & 0 & 1 & \\ & & & 0 & \ddots \end{pmatrix}.$$



Summary on Methods 1 & 2

- Both methods give **upper bounds** on $\text{spec } A$ and $\text{spec}_\varepsilon A$.

Summary on Methods 1 & 2

- Both methods give **upper bounds** on $\text{spec } A$ and $\text{spec}_\epsilon A$.
- The bound from Method 2 is often **sharper**.

Summary on Methods 1 & 2

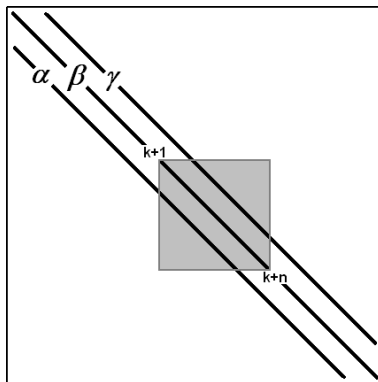
- Both methods give **upper bounds** on $\text{spec } A$ and $\text{spec}_\varepsilon A$.
- The bound from Method 2 is often **sharper**.
- Conjecture: Method 2 **converges** to $\text{spec}_\varepsilon A$ as $n \rightarrow \infty$.

Summary on Methods 1 & 2

- Both methods give **upper bounds** on $\text{spec } A$ and $\text{spec}_\epsilon A$.
- The bound from Method 2 is often **sharper**.
- Conjecture: Method 2 **converges** to $\text{spec}_\epsilon A$ as $n \rightarrow \infty$.
- Method 1 also works for **semi-infinite** and **finite** matrices A !

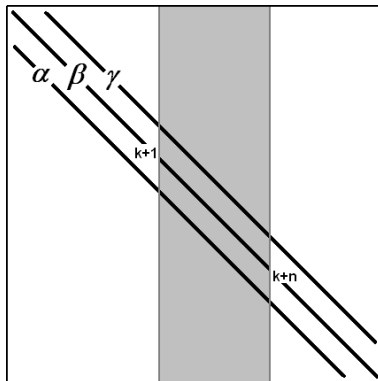
Here is another idea: Method 3

Instead of



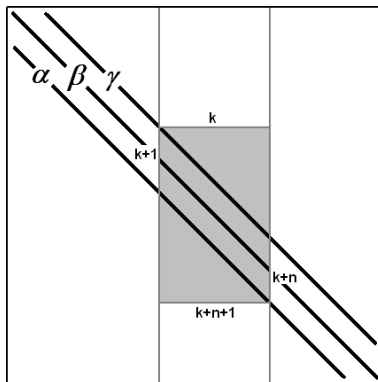
Here is another idea: Method 3

We do a “**one-sided**” truncation.



Here is another idea: Method 3

We do a “**one-sided**” truncation.



In a sense, we work with **rectangular** finite submatrices.

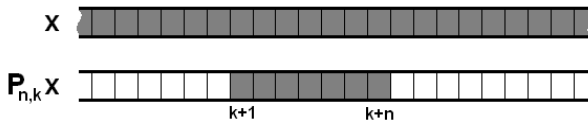
This is motivated by work of Davies 1998 and Hansen 2008.

(Also see Heinemeyer/Lindner/Potthast [SIAM Num. Anal. 2007].)

Method 3: Projection Operator

For $n \in \mathbb{N}$ and $k \in \mathbb{Z}$, let $P_{n,k} : \ell^2 \rightarrow \ell^2$ denote the projection

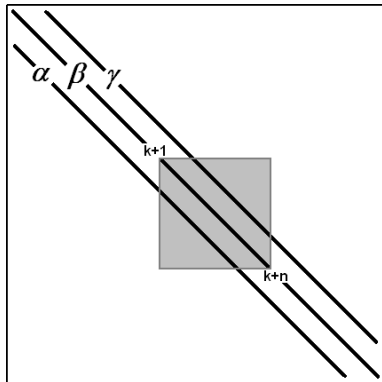
$$(P_{n,k}x)(i) := \begin{cases} x(i), & i \in \{k+1, \dots, k+n\}, \\ 0 & \text{otherwise.} \end{cases}$$



Further, we put $X_{n,k} := \text{im } P_{n,k}$ and identify it with \mathbb{C}^n in the obvious way.

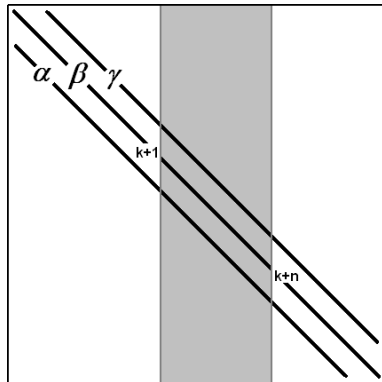
Method 3: Truncations

Method 1:



$$P_{n,k}(A - \lambda I)P_{n,k}|_{X_{n,k}}$$

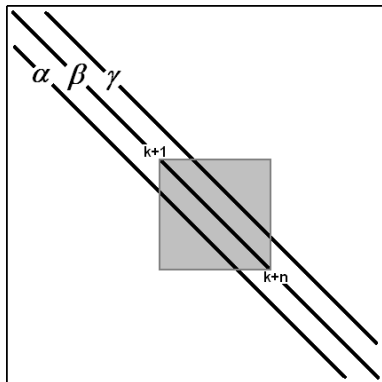
Method 3:



$$(A - \lambda I)P_{n,k}|_{X_{n,k}}$$

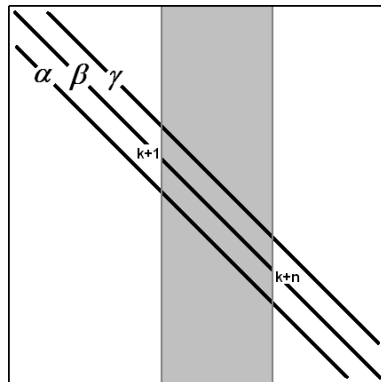
Method 3: Truncations

Method 1:



$$\mathbf{P}_{n,k}(A - \lambda I)\mathbf{P}_{n,k}|_{X_{n,k}}$$

Method 3:



$$(A - \lambda I)\mathbf{P}_{n,k}|_{X_{n,k}}$$

Method 1:

$\lambda \in \text{spec}_\varepsilon(A) \implies$ For some $k \in \mathbb{Z}$:

$$\lambda \in \text{spec}_{\varepsilon+\varepsilon_n}(P_{n,k}AP_{n,k}|_{X_{n,k}})$$

Method 1:

$\lambda \in \text{spec}_\varepsilon(A) \implies$ For some $k \in \mathbb{Z}$:

$$\lambda \in \text{spec}_{\varepsilon+\varepsilon_n}(P_{n,k}AP_{n,k}|_{X_{n,k}})$$

$$\text{i.e. } s_{\min}(P_{n,k}(A - \lambda I)P_{n,k}) < \varepsilon + \varepsilon_n$$

Method 1:

$\lambda \in \text{spec}_\varepsilon(A) \implies$ For some $k \in \mathbb{Z}$:

$$\lambda \in \text{spec}_{\varepsilon+\varepsilon_n}(P_{n,k}AP_{n,k}|_{X_{n,k}})$$

$$\text{i.e. } s_{\min}(P_{n,k}(A - \lambda I)P_{n,k}) < \varepsilon + \varepsilon_n$$

$$\min \text{spec} \left((P_{n,k}(A - \lambda I)P_{n,k})^* (P_{n,k}(A - \lambda I)P_{n,k}) \right) < (\varepsilon + \varepsilon_n)^2$$

Method 1:

$\lambda \in \text{spec}_\varepsilon(A) \implies$ For some $k \in \mathbb{Z}$:

$$\lambda \in \text{spec}_{\varepsilon+\varepsilon_n}(P_{n,k}AP_{n,k}|_{X_{n,k}})$$

$$\text{i.e. } s_{\min}(P_{n,k}(A - \lambda I)P_{n,k}) < \varepsilon + \varepsilon_n$$

$$\min \text{spec} \left((P_{n,k}(A - \lambda I)^*P_{n,k})(P_{n,k}(A - \lambda I)P_{n,k}) \right) < (\varepsilon + \varepsilon_n)^2$$

Method 1:

$\lambda \in \text{spec}_\varepsilon(A) \implies$ For some $k \in \mathbb{Z}$:

$$\lambda \in \text{spec}_{\varepsilon+\varepsilon_n}(P_{n,k}AP_{n,k}|_{X_{n,k}})$$

$$\text{i.e. } s_{\min}(P_{n,k}(A - \lambda I)P_{n,k}) < \varepsilon + \varepsilon_n$$

$$\min \text{spec} \left(P_{n,k}(A - \lambda I)^*P_{n,k}P_{n,k}(A - \lambda I)P_{n,k} \right) < (\varepsilon + \varepsilon_n)^2$$

Method 1:

$\lambda \in \text{spec}_\varepsilon(A) \implies$ For some $k \in \mathbb{Z}$:

$$\lambda \in \text{spec}_{\varepsilon+\varepsilon_n}(P_{n,k}AP_{n,k}|_{X_{n,k}})$$

$$\text{i.e. } s_{\min}(P_{n,k}(A - \lambda I)P_{n,k}) < \varepsilon + \varepsilon_n$$

$$\min \text{spec} \left(P_{n,k}(A - \lambda I)^*P_{n,k}(A - \lambda I)P_{n,k} \right) < (\varepsilon + \varepsilon_n)^2$$

Method 1:

$\lambda \in \text{spec}_\varepsilon(A) \implies$ For some $k \in \mathbb{Z}$:

$$\lambda \in \text{spec}_{\varepsilon+\varepsilon_n}(P_{n,k}AP_{n,k}|_{X_{n,k}})$$

$$\text{i.e. } s_{\min}(P_{n,k}(A - \lambda I)P_{n,k}) < \varepsilon + \varepsilon_n$$

Idea: $\min \text{spec} \left(P_{n,k}(A - \lambda I)^* \cancel{P_{n,k}}(A - \lambda I)P_{n,k} \right) < (\varepsilon + \varepsilon_n)^2$

Method 1:

$\lambda \in \text{spec}_\varepsilon(A) \implies$ For some $k \in \mathbb{Z}$:

$$\lambda \in \text{spec}_{\varepsilon+\varepsilon_n}(P_{n,k}AP_{n,k}|_{X_{n,k}})$$

$$\text{i.e. } s_{\min}(\cancel{P_{n,k}}(A - \lambda I)P_{n,k}) < \varepsilon + \varepsilon_n$$

Idea: $\min \text{spec} \left(P_{n,k}(A - \lambda I)^* \cancel{P_{n,k}}(A - \lambda I)P_{n,k} \right) < (\varepsilon + \varepsilon_n)^2$

Method 1:

$\lambda \in \text{spec}_\varepsilon(A) \implies$ For some $k \in \mathbb{Z}$:

$$\lambda \in \text{spec}_{\varepsilon+\varepsilon_n}(\cancel{P_{n,k}}AP_{n,k}|_{X_{n,k}})$$

$$\text{i.e. } s_{\min}(\cancel{P_{n,k}}(A - \lambda I)P_{n,k}) < \varepsilon + \varepsilon_n$$

Idea: $\min \text{spec} \left(P_{n,k}(A - \lambda I)^* \cancel{P_{n,k}}(A - \lambda I)P_{n,k} \right) < (\varepsilon + \varepsilon_n)^2$

Method 1:

Idea: $\min \operatorname{spec} \left(P_{n,k}(A - \lambda I)^* \cancel{P_{n,k}}(A - \lambda I)P_{n,k} \right) < (\varepsilon + \varepsilon_n)^2$

Method 1:

Idea: $\min \operatorname{spec} \left(P_{n,k}(A - \lambda I)^* \cancel{P_{n,k}}(A - \lambda I)P_{n,k} \right) < (\varepsilon + \varepsilon_n)^2$

Method 3

Let $\gamma_\varepsilon^{n,k}(A)$ be the set of all $\lambda \in \mathbb{C}$, for which

$$\min \operatorname{spec} \left(P_{n,k}(A - \lambda I)^*(A - \lambda I)P_{n,k} \right) < (\varepsilon + \varepsilon_n)^2$$

Method 1:

Idea: $\min \text{spec} \left(P_{n,k}(A - \lambda I)^* \cancel{P_{n,k}}(A - \lambda I)P_{n,k} \right) < (\varepsilon + \varepsilon_n)^2$

Method 3

Let $\gamma_\varepsilon^{n,k}(A)$ be the set of all $\lambda \in \mathbb{C}$, for which

$$\min \text{spec} \left(P_{n,k}(A - \lambda I)^*(A - \lambda I)P_{n,k} \right) < (\varepsilon + \varepsilon_n)^2$$

and $\min \text{spec} \left(P_{n,k}(A - \lambda I)(A - \lambda I)^*P_{n,k} \right) < (\varepsilon + \varepsilon_n)^2.$

Method 1:

Idea: $\min \operatorname{spec} \left(P_{n,k}(A - \lambda I)^* \cancel{P_{n,k}} (A - \lambda I) P_{n,k} \right) < (\varepsilon + \varepsilon_n)^2$

Method 3

Let $\gamma_\varepsilon^{n,k}(A)$ be the set of all $\lambda \in \mathbb{C}$, for which

$$\min \operatorname{spec} \left(P_{n,k}(A - \lambda I)^*(A - \lambda I)P_{n,k} \right) < (\varepsilon + \varepsilon_n)^2$$

and $\min \operatorname{spec} \left(P_{n,k}(A - \lambda I)(A - \lambda I)^*P_{n,k} \right) < (\varepsilon + \varepsilon_n)^2.$

Then put

$$\Gamma_\varepsilon^n(A) := \bigcup_{k \in \mathbb{Z}} \gamma_\varepsilon^{n,k}(A).$$

Method 3: Spectral bounds

Again we get (as in Methods 1 & 2)

Upper Bound

$$\text{spec}_\varepsilon(A) \subset \bigcup_{k \in \mathbb{Z}} \gamma_{\varepsilon + \varepsilon_n}^{n,k}(A) = \Gamma_{\varepsilon + \varepsilon_n}^n(A)$$

with $\varepsilon_n = 2(\|\alpha\|_\infty + \|\gamma\|_\infty) \sin \frac{\pi}{2n+2} < \frac{\pi}{n+1}(\|\alpha\|_\infty + \|\gamma\|_\infty)$

and this time the upper bound looks even sharper than before.

Method 3: Spectral bounds

Again we get (as in Methods 1 & 2)

Upper Bound

$$\operatorname{spec}_\varepsilon(A) \subset \bigcup_{k \in \mathbb{Z}} \gamma_{\varepsilon + \varepsilon_n}^{n,k}(A) = \Gamma_{\varepsilon + \varepsilon_n}^n(A)$$

with $\varepsilon_n = 2(\|\alpha\|_\infty + \|\gamma\|_\infty) \sin \frac{\pi}{2n+2} < \frac{\pi}{n+1}(\|\alpha\|_\infty + \|\gamma\|_\infty)$

and this time the upper bound looks even sharper than before.
But now we also have

Lower Bound

$$\Gamma_\varepsilon^n(A) \subset \operatorname{spec}_\varepsilon(A).$$

Method 3: Spectral bounds

From the lower and upper bound

$$\Gamma_{\varepsilon}^n(A) \subset \operatorname{spec}_{\varepsilon}(A) \quad \text{and} \quad \operatorname{spec}_{\varepsilon}(A) \subset \Gamma_{\varepsilon+\varepsilon_n}^n(A)$$

we get

Sandwich 1

$$\Gamma_{\varepsilon}^n(A) \subset \operatorname{spec}_{\varepsilon}(A) \subset \Gamma_{\varepsilon+\varepsilon_n}^n(A)$$

Method 3: Spectral bounds

From the lower and upper bound

$$\Gamma_{\varepsilon}^n(A) \subset \operatorname{spec}_{\varepsilon}(A) \quad \text{and} \quad \operatorname{spec}_{\varepsilon}(A) \subset \Gamma_{\varepsilon+\varepsilon_n}^n(A)$$

we get

Sandwich 1

$$\Gamma_{\varepsilon}^n(A) \subset \operatorname{spec}_{\varepsilon}(A) \subset \Gamma_{\varepsilon+\varepsilon_n}^n(A)$$

Sandwich 2

$$\operatorname{spec}_{\varepsilon}(A) \subset \Gamma_{\varepsilon+\varepsilon_n}^n(A) \subset \operatorname{spec}_{\varepsilon+\varepsilon_n}(A).$$

Method 3: Spectral bounds

From the lower and upper bound

$$\Gamma_{\varepsilon}^n(A) \subset \operatorname{spec}_{\varepsilon}(A) \quad \text{and} \quad \operatorname{spec}_{\varepsilon}(A) \subset \Gamma_{\varepsilon+\varepsilon_n}^n(A)$$

we get

Sandwich 1

$$\Gamma_{\varepsilon}^n(A) \subset \operatorname{spec}_{\varepsilon}(A) \subset \Gamma_{\varepsilon+\varepsilon_n}^n(A)$$

Sandwich 2

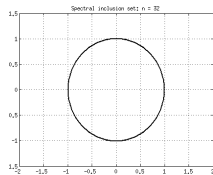
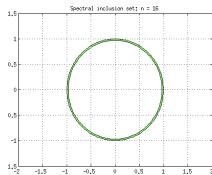
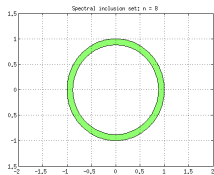
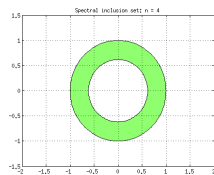
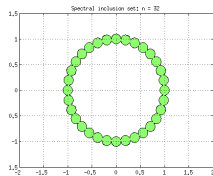
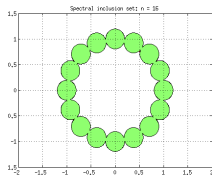
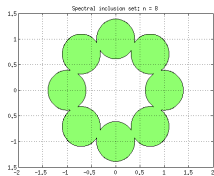
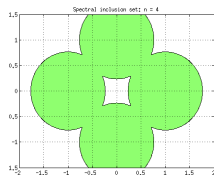
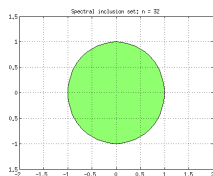
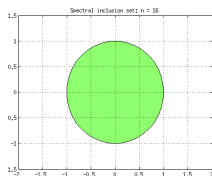
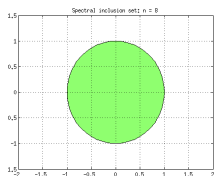
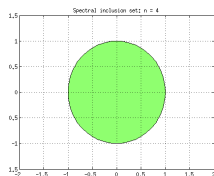
$$\operatorname{spec}_{\varepsilon}(A) \subset \Gamma_{\varepsilon+\varepsilon_n}^n(A) \subset \operatorname{spec}_{\varepsilon+\varepsilon_n}(A).$$

In particular, it follows that

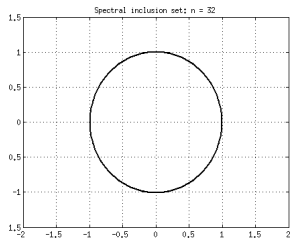
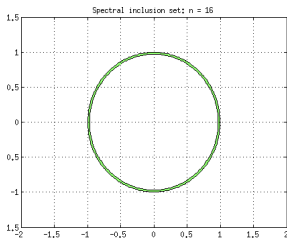
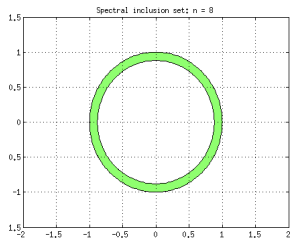
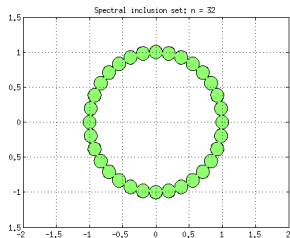
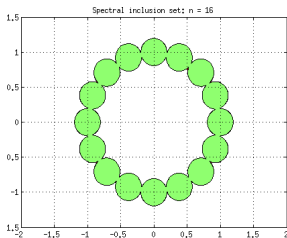
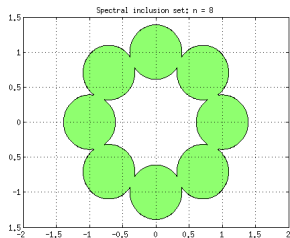
$$\Gamma_{\varepsilon+\varepsilon_n}^n(A) \rightarrow \operatorname{spec}_{\varepsilon}(A), \quad n \rightarrow \infty$$

in the Hausdorff metric.

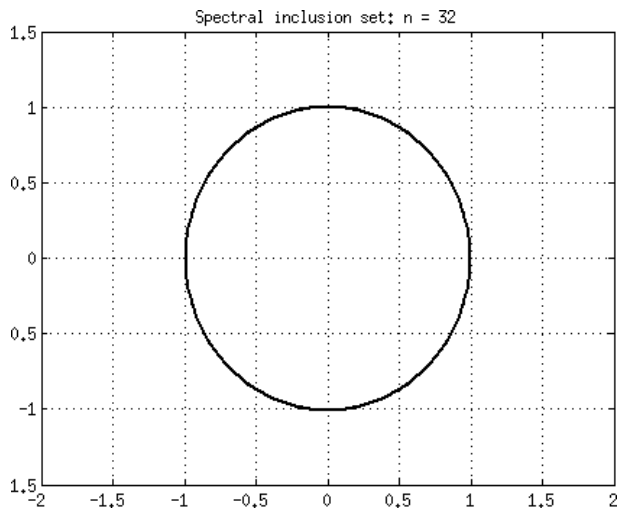
Methods 1, 2 & 3: The Shift Operator



Methods 2 & 3: The Shift Operator



Method 3: The Shift Operator

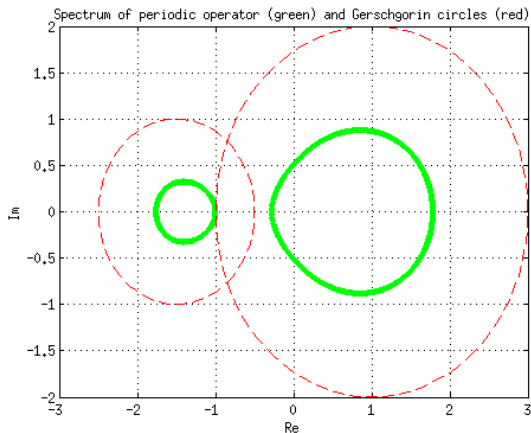


Methods 1, 2 & 3: Second Example

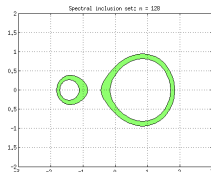
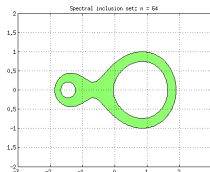
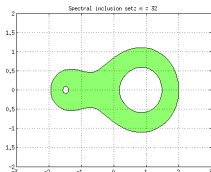
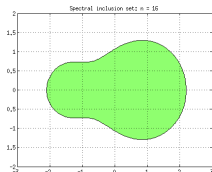
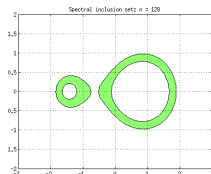
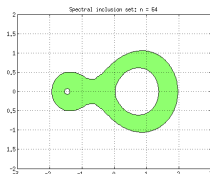
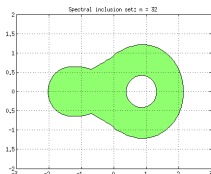
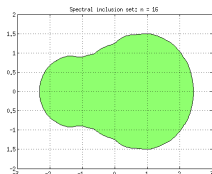
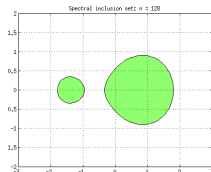
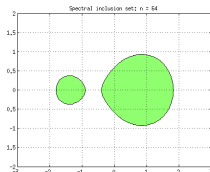
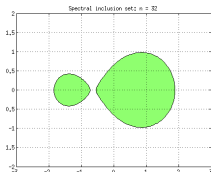
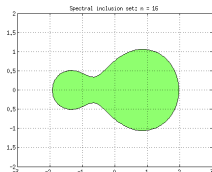
We now look at a matrix A with 3-periodic diagonals:

main diagonal: $\dots, -\frac{3}{2}, 1, 1, \dots$

super-diagonal: $\dots, 1, 2, 1, \dots$



Methods 1, 2 & 3: Second Example

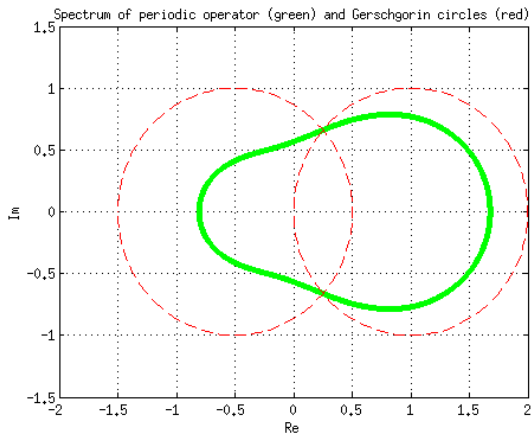


Methods 1, 2 & 3: Third Example

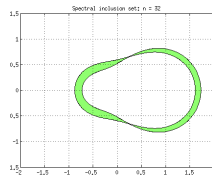
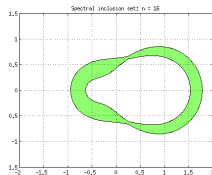
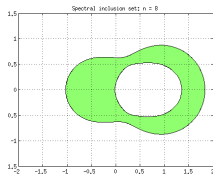
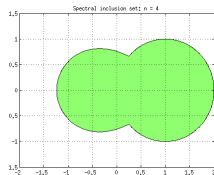
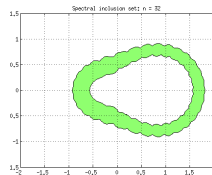
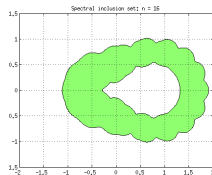
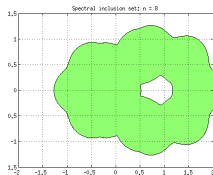
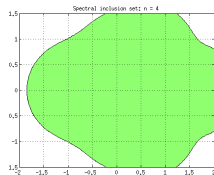
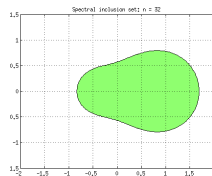
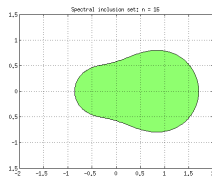
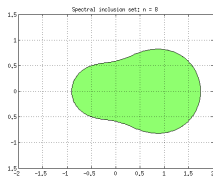
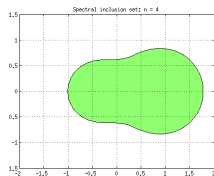
We now look at a matrix A with 3-periodic diagonals:

main diagonal: $\dots, -\frac{1}{2}, 1, 1, \dots$

super-diagonal: $\dots, 1, 1, 1, \dots$



Methods 1, 2 & 3: Third Example



- 1 Introduction
- 2 Inclusion Sets and Approximations for the Spectrum and Pseudospectrum
- 3 Substantial example with lots of pictures! Feinberg-Zee Random Hopping Matrix
- 4 Extensions

From a talk of Anthony Zee (MSRI Berkeley, 1999)

Exploring Non-Hermitian Random Matrices¹⁶
largely terra incognita !

Consider

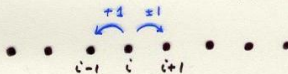
$$H = \begin{pmatrix} 0 & \neq 1 & & & \\ 1 & 0 & \neq 1 & & \\ & 1 & 0 & \neq 1 & \\ & & 1 & & \\ & & & & \ddots \\ \textcircled{\pm 1} & & & & & \textcircled{1} \\ & & & & & & \ddots \\ & & & & & & & \textcircled{\pm 1} \\ & & & & & & & 1 & 0 \end{pmatrix}$$

+1 or -1
with probability $\frac{1}{2}$

If H regarded as Hamiltonian of quantum particle
 $\sum_j H_{ij} \psi_j = E \psi_i$

$$\Rightarrow \psi_{i-1} + t_i \psi_{i+1} = E \psi_i, \quad t_i = \pm 1$$

Feynberg & Zee.
Nucl. Phys



What does the spectrum look like ?

Look at the **bi-infinite** matrix

$$A^b = \begin{pmatrix} \ddots & & & & \\ & \ddots & & & \\ & & 0 & 1 & \\ & & b_{-1} & 0 & 1 \\ & & & b_0 & 0 & 1 \\ & & & & b_1 & 0 & \ddots \\ & & & & & \ddots & \ddots \end{pmatrix},$$

where $b = (\cdots, b_{-1}, b_0, b_1, \cdots) \in \{\pm 1\}^{\mathbb{Z}}$ is a **pseudoergodic** sequence

Look at the **bi-infinite** matrix

$$A^b = \begin{pmatrix} \ddots & & & & \\ & \ddots & & & \\ & & 0 & 1 & \\ & & b_{-1} & 0 & 1 \\ & & & b_0 & 0 & 1 \\ & & & & b_1 & 0 & \ddots \\ & & & & & \ddots & \ddots \end{pmatrix},$$

where $b = (\cdots, b_{-1}, b_0, b_1, \cdots) \in \{\pm 1\}^{\mathbb{Z}}$ is a **pseudoergodic** sequence; that means:

every finite pattern of ± 1 's can be found somewhere in the infinite sequence b .

Spectral Formula

C-W, Lindner 2008

If b is pseudoergodic then

$$\operatorname{spec} A^b = \operatorname{spec}_{\text{ess}} A^b = \bigcup_{c \in \{\pm 1\}^{\mathbb{Z}}} \operatorname{spec}_{\text{point}}^{\infty} A^c.$$

(Also see Rabinovich/Roch/Silbermann 1998 and Davies 2001.)

Spectral Formula

C-W, Lindner 2008

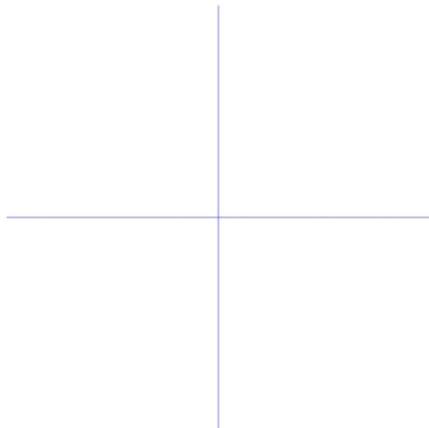
If b is pseudoergodic then

$$\operatorname{spec} A^b = \operatorname{spec}_{\text{ess}} A^b = \bigcup_{c \in \{\pm 1\}^{\mathbb{Z}}} \operatorname{spec}_{\text{point}}^{\infty} A^c.$$

(Also see Rabinovich/Roch/Silbermann 1998 and Davies 2001.)

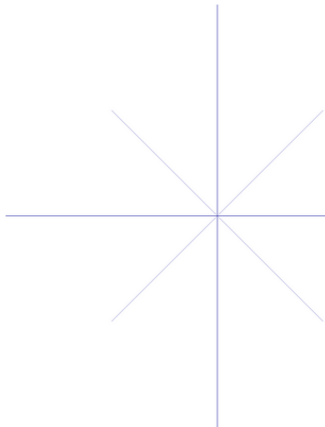
One can try to “exhaust” the RHS by running through all **periodic** ± 1 sequences c .

Final Example [Feinberg/Zee 1999]

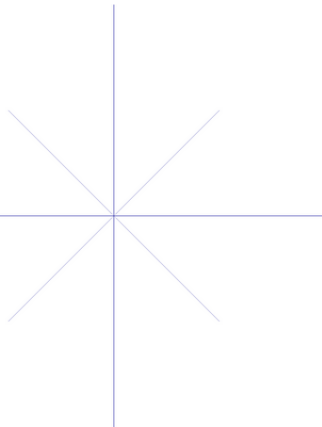


Period 1

Final Example [Feinberg/Zee 1999]

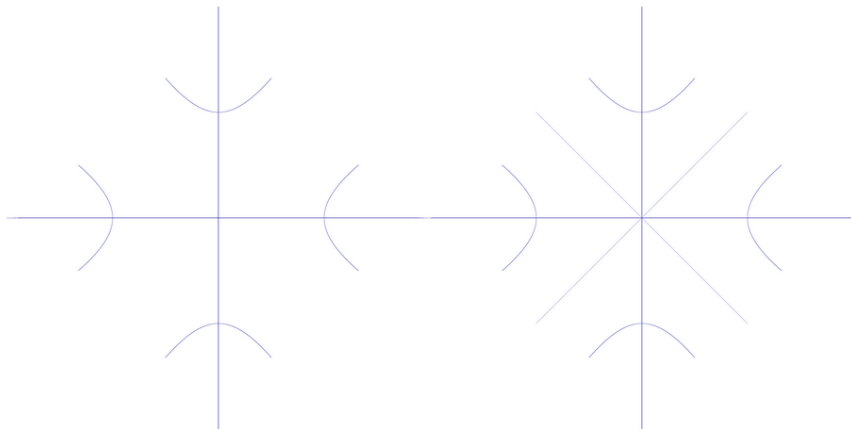


Period 2



Periods 1, 2

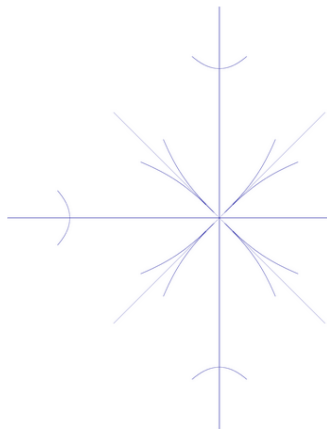
Final Example [Feinberg/Zee 1999]



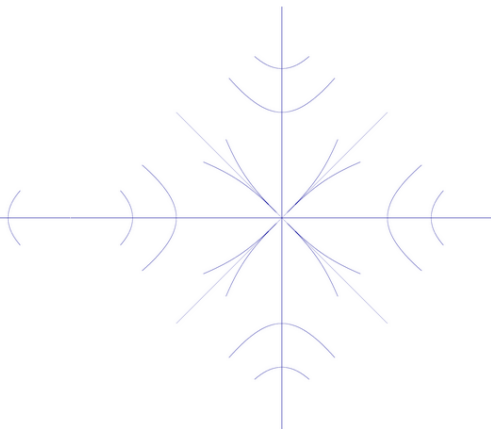
Period 3

Periods 1, ..., 3

Final Example [Feinberg/Zee 1999]

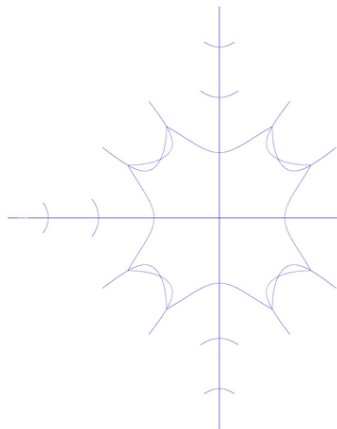


Period 4

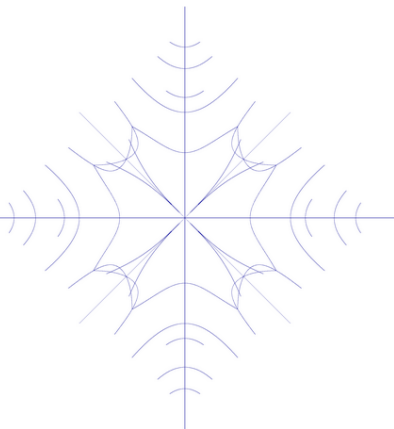


Periods 1, ..., 4

Final Example [Feinberg/Zee 1999]

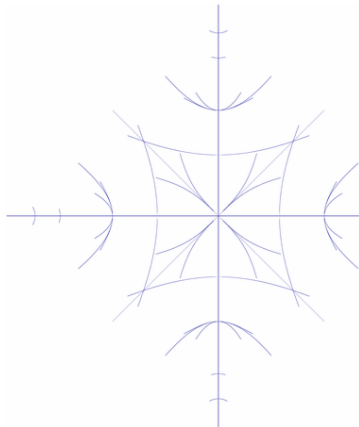


Period 5

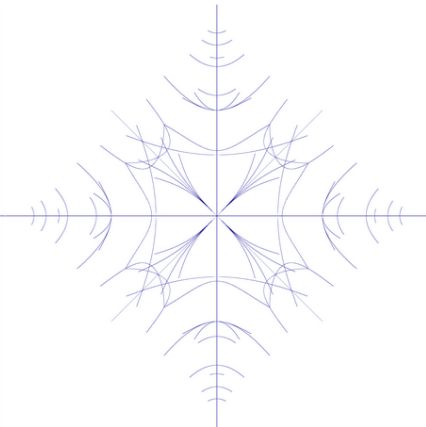


Periods 1, ..., 5

Final Example [Feinberg/Zee 1999]

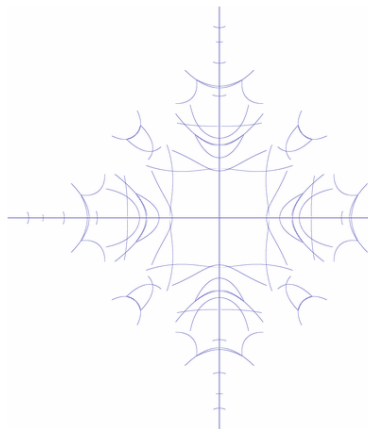


Period 6

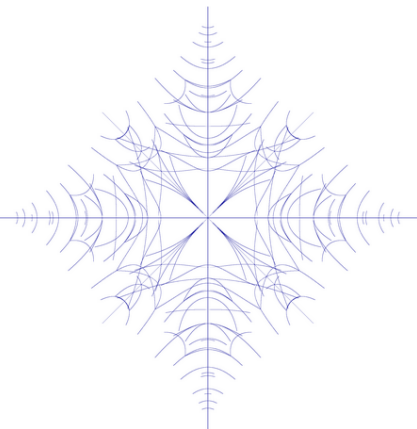


Periods 1, ..., 6

Final Example [Feinberg/Zee 1999]

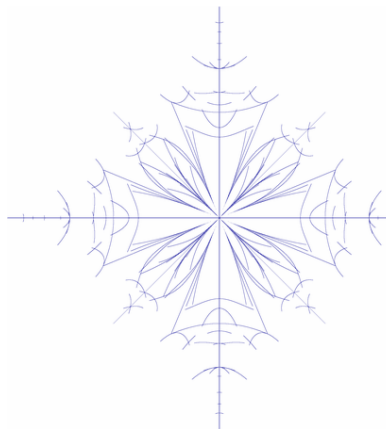


Period 7

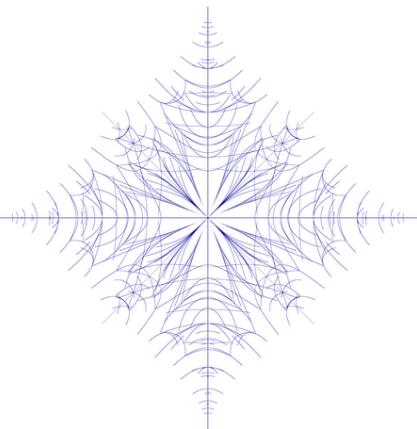


Periods 1, ..., 7

Final Example [Feinberg/Zee 1999]

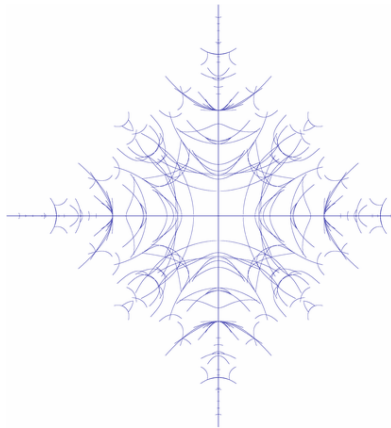


Period 8

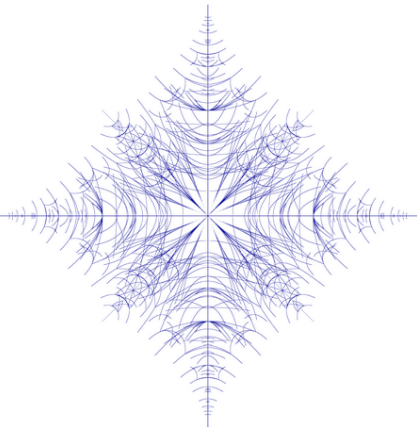


Periods 1, ..., 8

Final Example [Feinberg/Zee 1999]

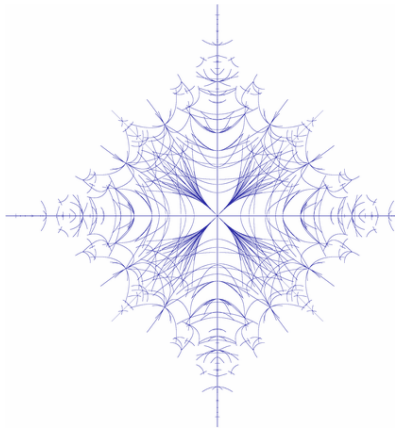


Period 9

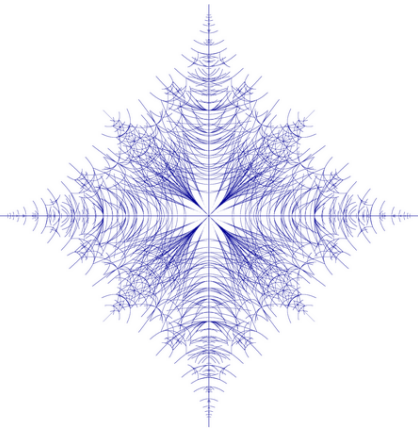


Periods 1, ..., 9

Final Example [Feinberg/Zee 1999]

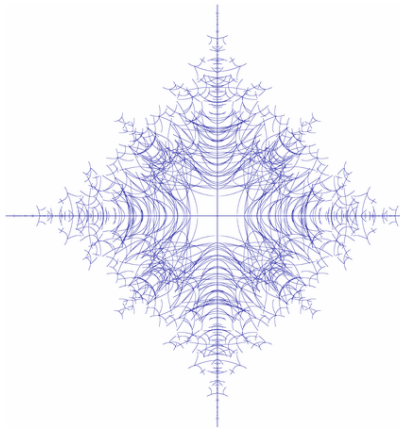


Period 10

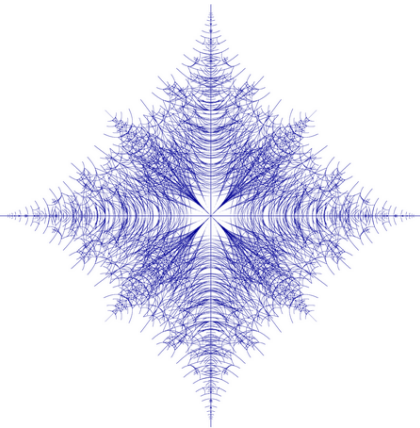


Periods 1, ..., 10

Final Example [Feinberg/Zee 1999]

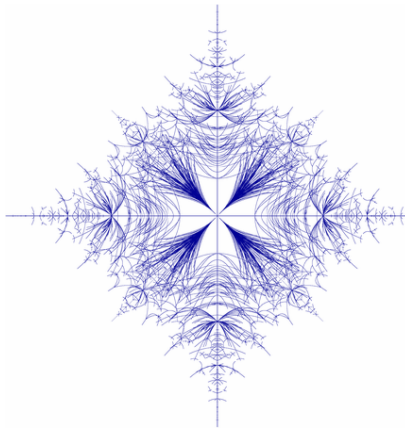


Period 11

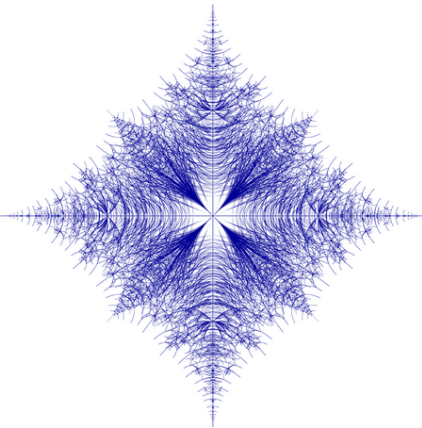


Periods 1, ..., 11

Final Example [Feinberg/Zee 1999]

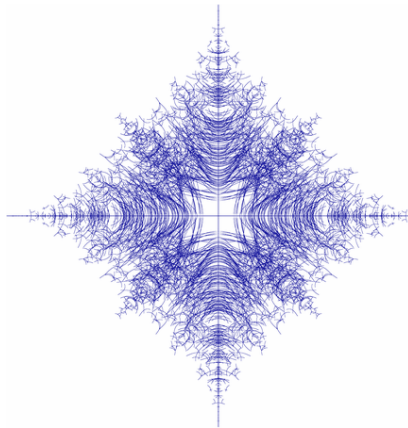


Period 12

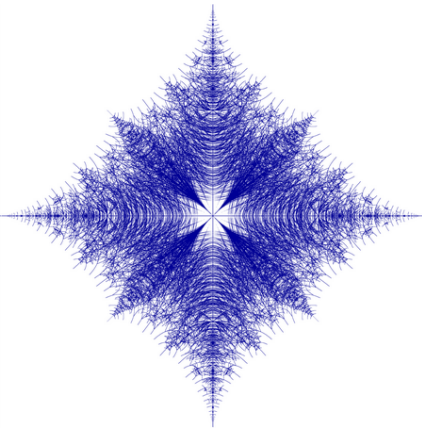


Periods 1, ..., 12

Final Example [Feinberg/Zee 1999]

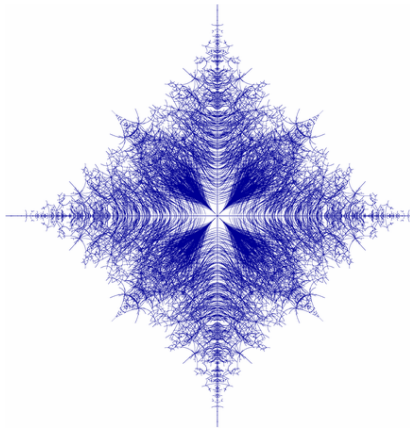


Period 13

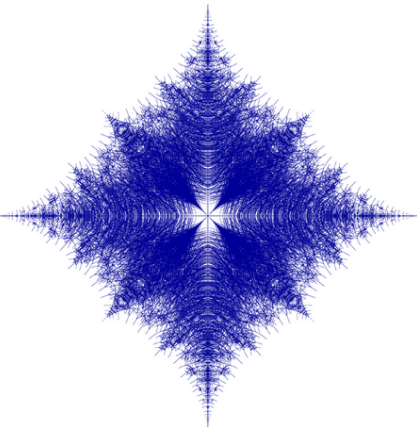


Periods 1, ..., 13

Final Example [Feinberg/Zee 1999]

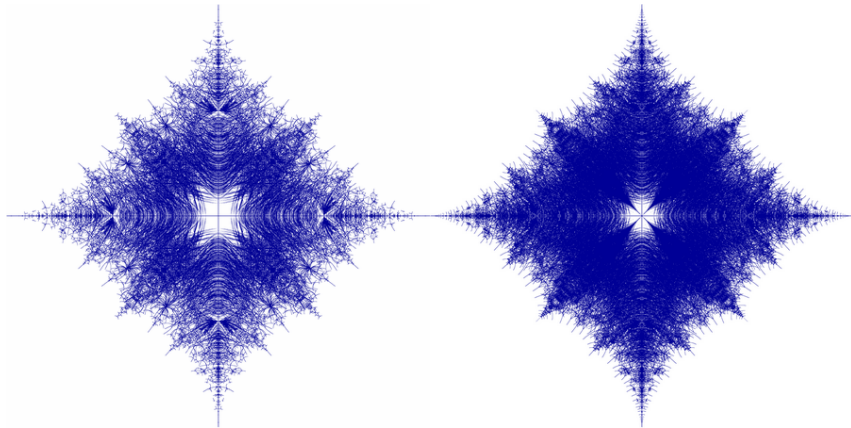


Period 14



Periods 1, ..., 14

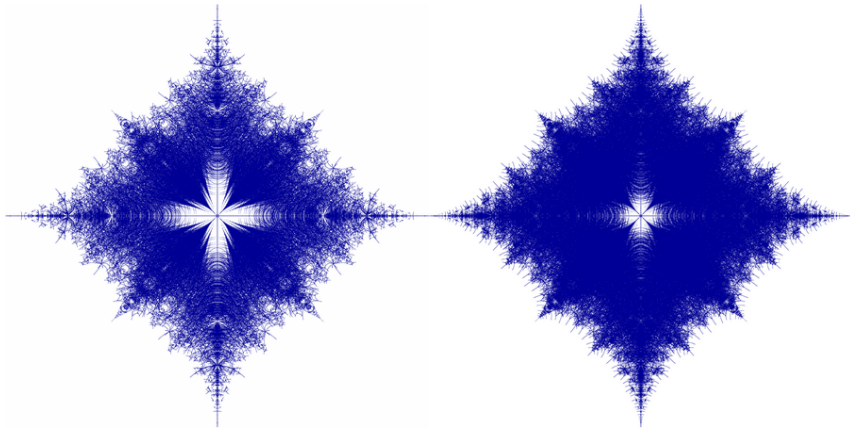
Final Example [Feinberg/Zee 1999]



Period 15

Periods 1, ..., 15

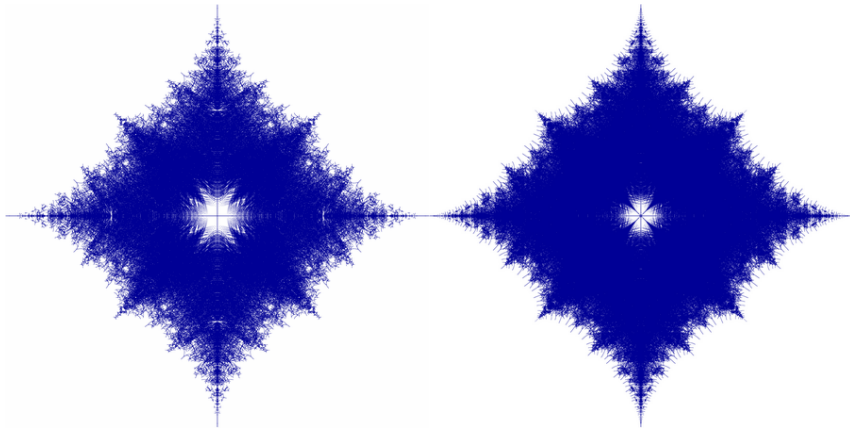
Final Example [Feinberg/Zee 1999]



Period 16

Periods 1, ..., 16

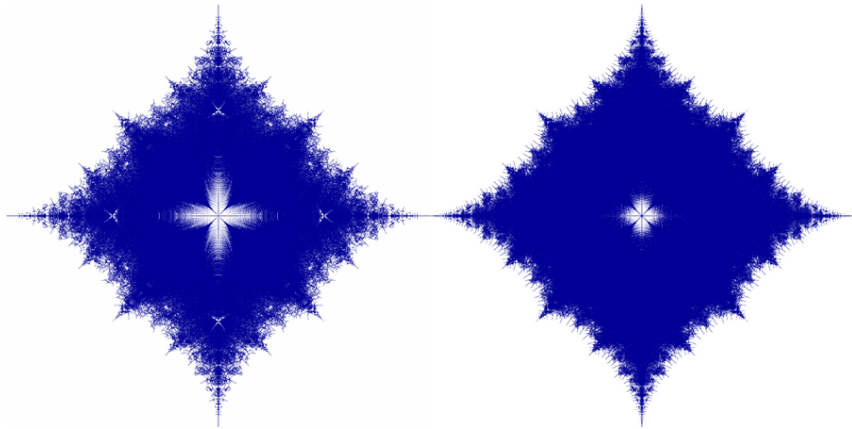
Final Example [Feinberg/Zee 1999]



Period 17

Periods 1, ..., 17

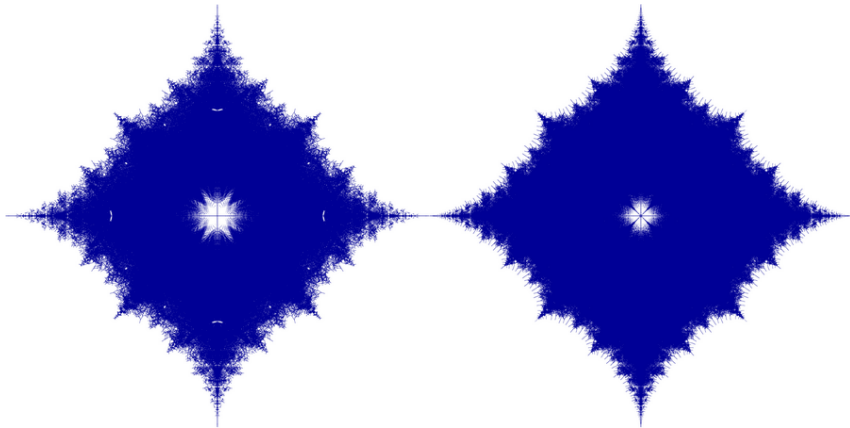
Final Example [Feinberg/Zee 1999]



Period 18

Periods 1, ..., 18

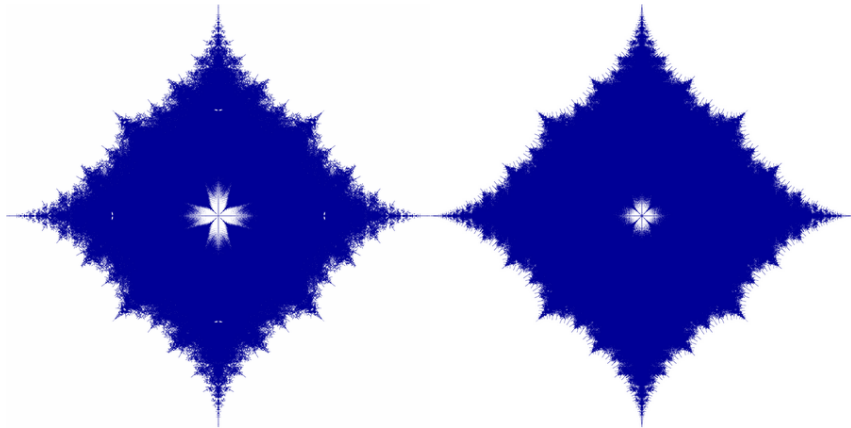
Final Example [Feinberg/Zee 1999]



Period 19

Periods 1, ..., 19

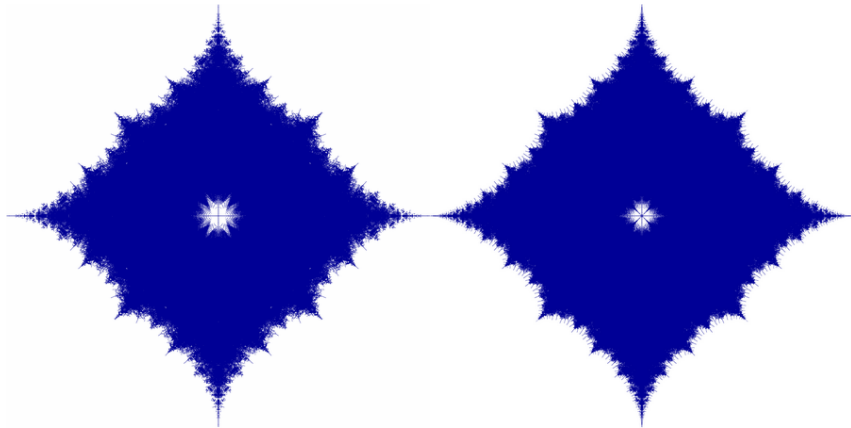
Final Example [Feinberg/Zee 1999]



Period 20

Periods 1, ..., 20

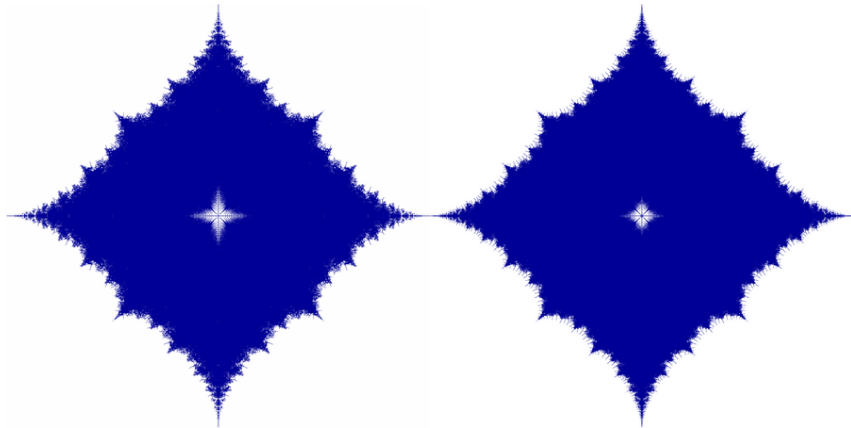
Final Example [Feinberg/Zee 1999]



Period 21

Periods 1, ..., 21

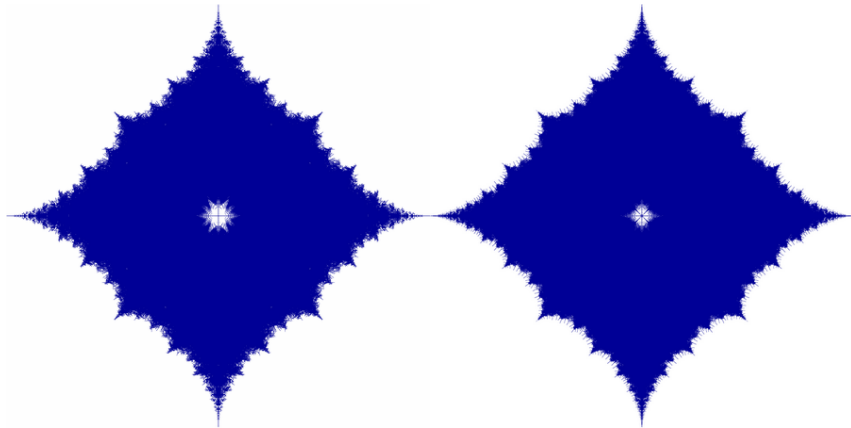
Final Example [Feinberg/Zee 1999]



Period 22

Periods 1, ..., 22

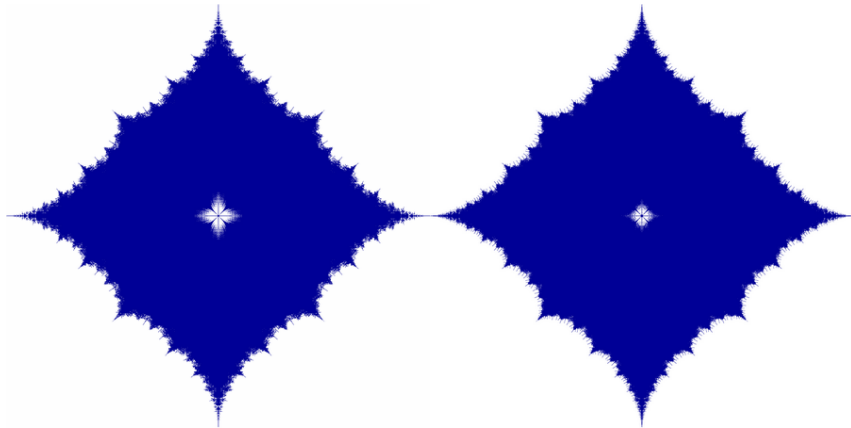
Final Example [Feinberg/Zee 1999]



Period 23

Periods 1, ..., 23

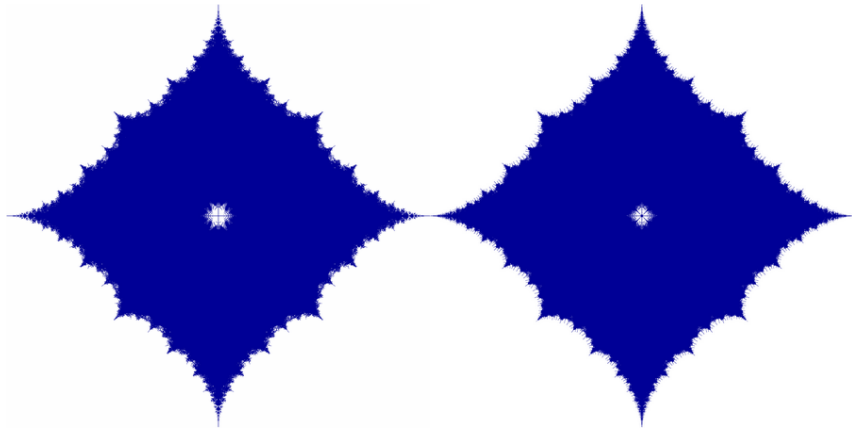
Final Example [Feinberg/Zee 1999]



Period 24

Periods 1, ..., 24

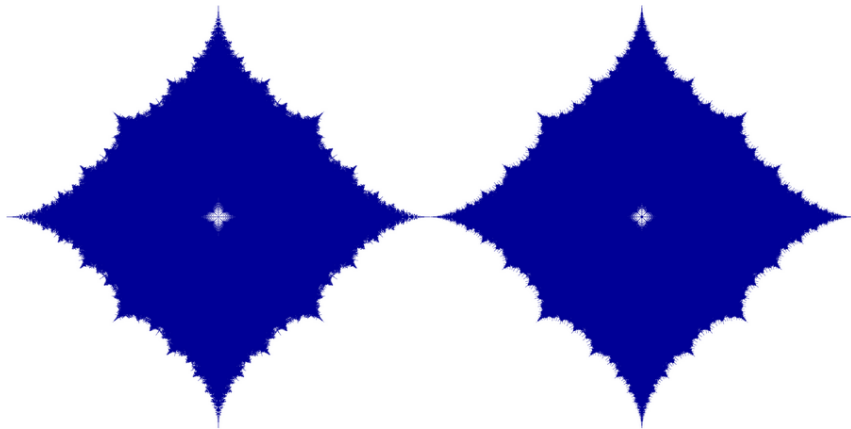
Final Example [Feinberg/Zee 1999]



Period 25

Periods 1, ..., 25

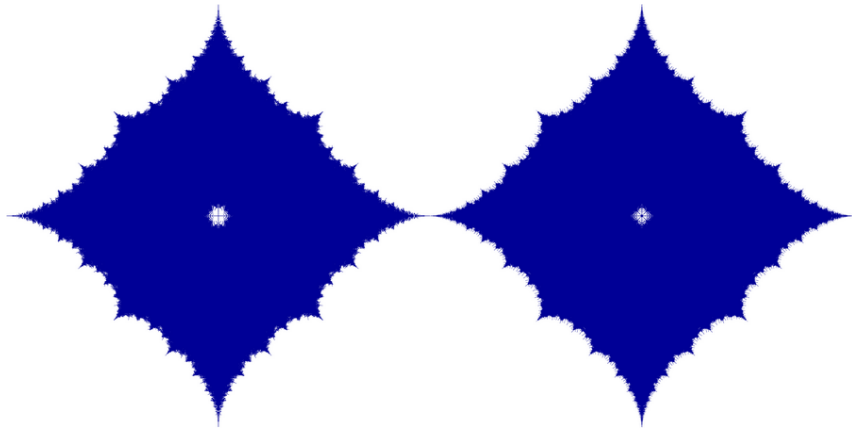
Final Example [Feinberg/Zee 1999]



Period 26

Periods 1, ..., 26

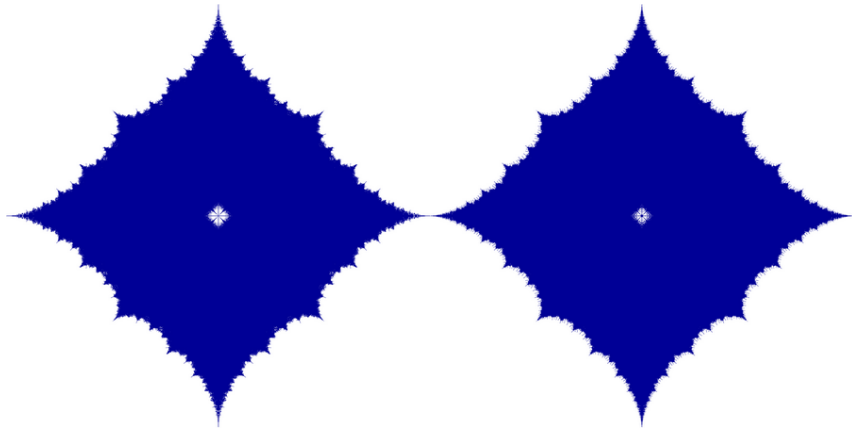
Final Example [Feinberg/Zee 1999]



Period 27

Periods 1, ..., 27

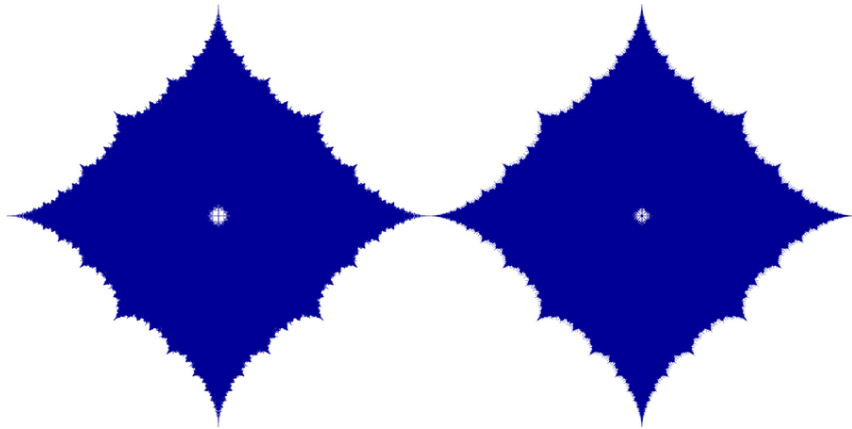
Final Example [Feinberg/Zee 1999]



Period 28

Periods 1, ..., 28

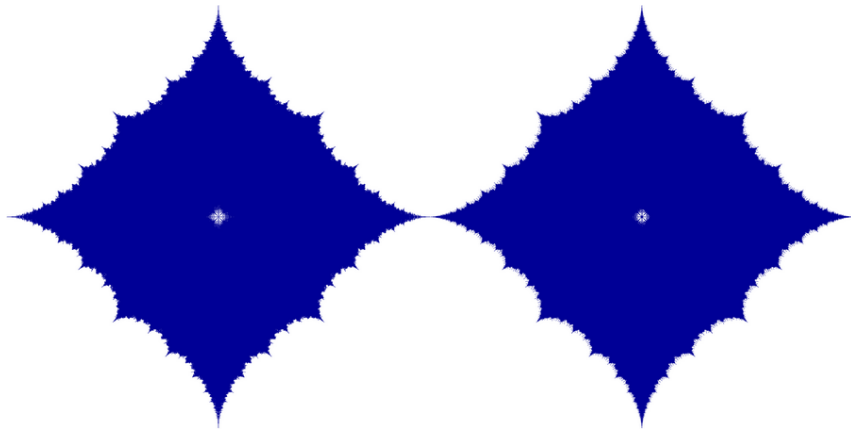
Final Example [Feinberg/Zee 1999]



Period 29

Periods 1, ..., 29

Final Example [Feinberg/Zee 1999]



Period 30

Periods 1, ..., 30

Recall our “Sandwich 1”: In this example, one has

$$\underbrace{\bigcup_{k \in \mathbb{Z}} \text{spec}_{\varepsilon}(P_{n,k} A^b P_{n,k})}_{=:\sigma_n^{\varepsilon}} \subset \text{spec}_{\varepsilon}(A^b) \subset \underbrace{\bigcup_{k \in \mathbb{Z}} \text{spec}_{\varepsilon + \varepsilon_n}(P_{n,k} A^b P_{n,k})}_{\Sigma_n^{\varepsilon} := \sigma_n^{\varepsilon + \varepsilon_n}}$$

for all $n \in \mathbb{N}$, so let's look at σ_n^{ε} for $\varepsilon = 0$.

Here are the $n \times n$ matrix eigenvalues

$$\sigma_n^0 = \bigcup_{k \in \mathbb{Z}} \text{spec}(P_{n,k} A^b P_{n,k})$$

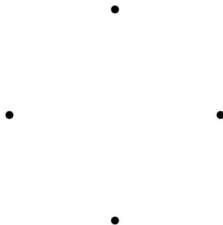
for $n = 1, \dots, 30$:

Final Example [Feinberg/Zee 1999]

•

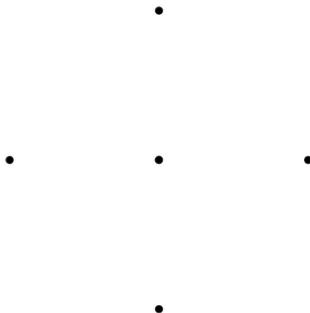
Size 1

Final Example [Feinberg/Zee 1999]



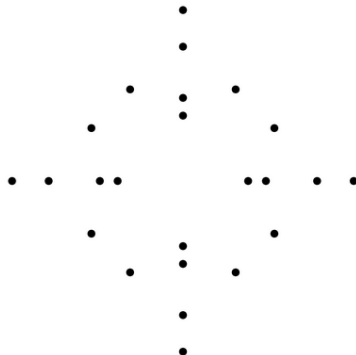
Size 2

Final Example [Feinberg/Zee 1999]



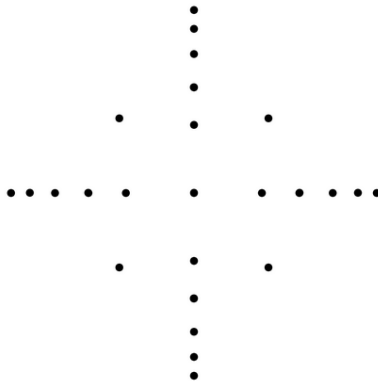
Size 3

Final Example [Feinberg/Zee 1999]



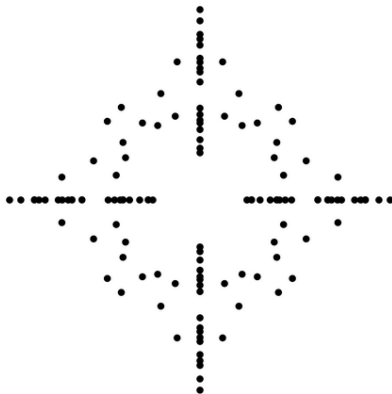
Size 4

Final Example [Feinberg/Zee 1999]



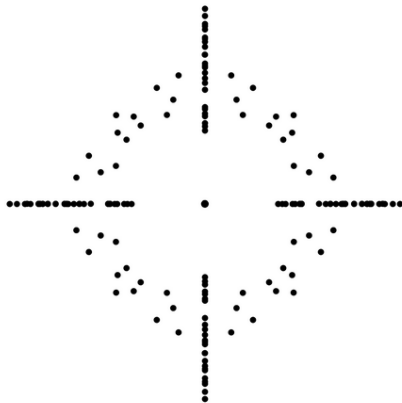
Size 5

Final Example [Feinberg/Zee 1999]



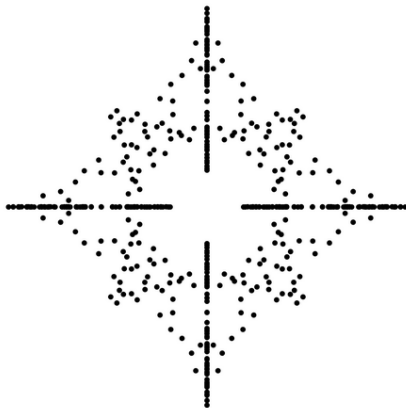
Size 6

Final Example [Feinberg/Zee 1999]



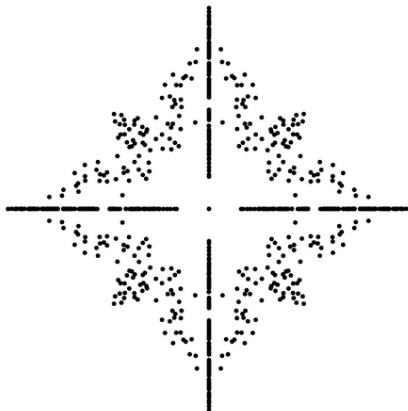
Size 7

Final Example [Feinberg/Zee 1999]



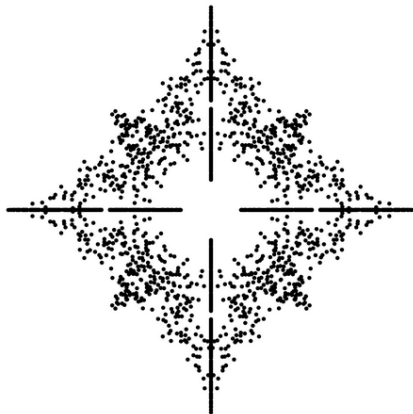
Size 8

Final Example [Feinberg/Zee 1999]

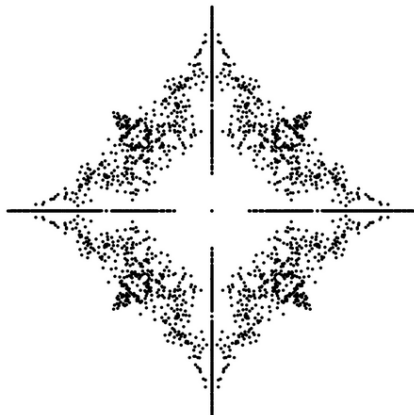


Size 9

Final Example [Feinberg/Zee 1999]

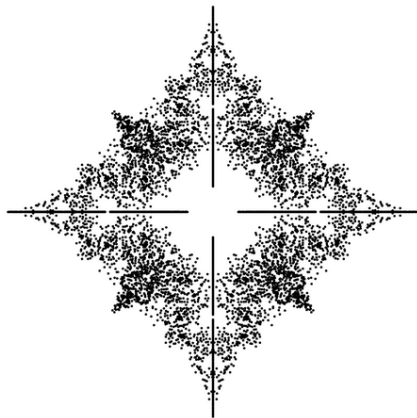


Size 10

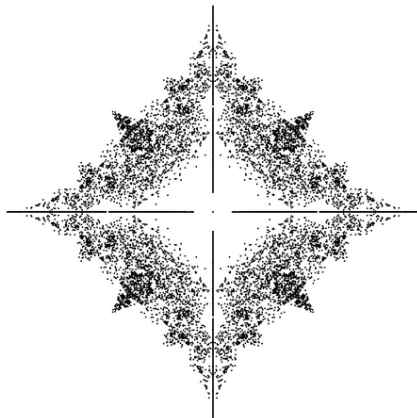


Size 11

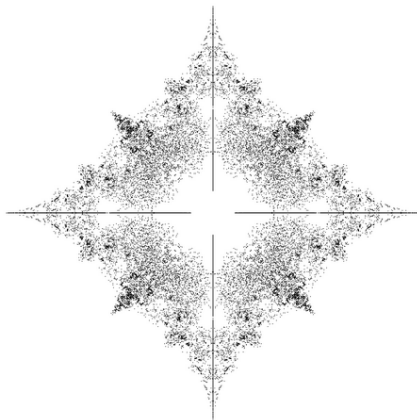
Final Example [Feinberg/Zee 1999]



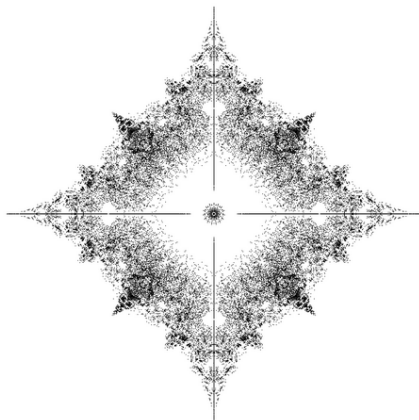
Size 12



Size 13

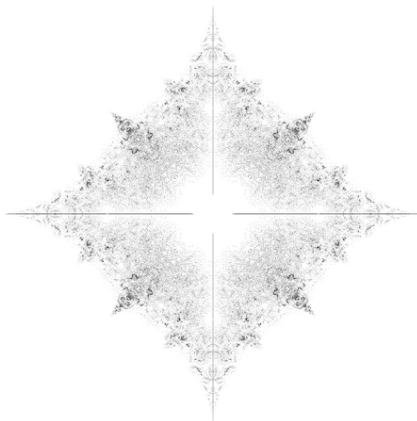


Size 14

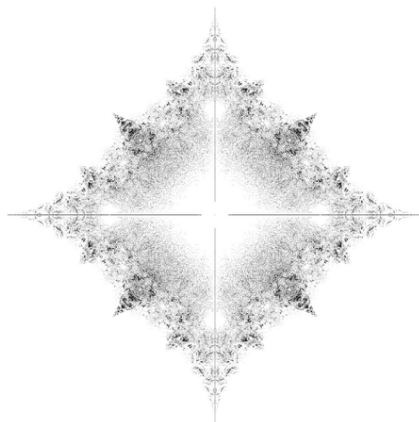


Size 15

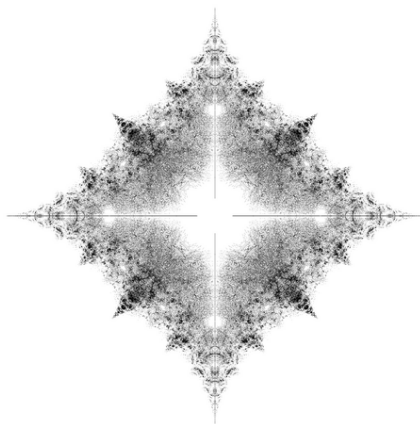
Final Example [Feinberg/Zee 1999]



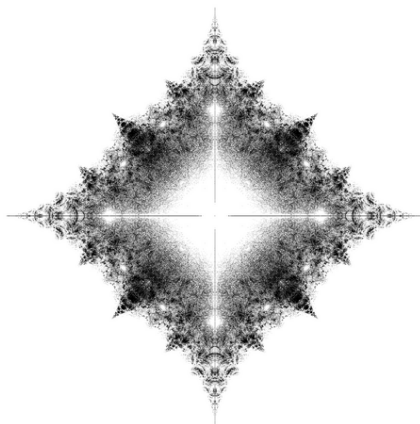
Size 16



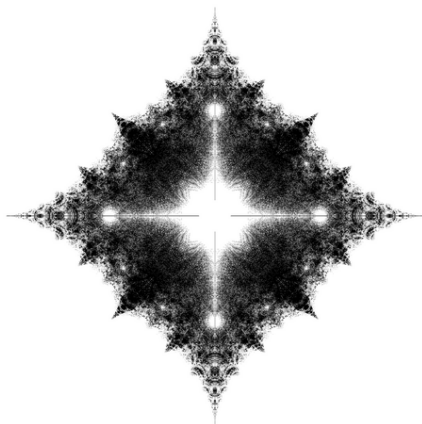
Size 17



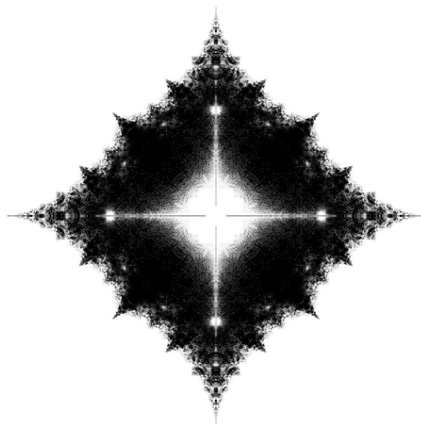
Size 18



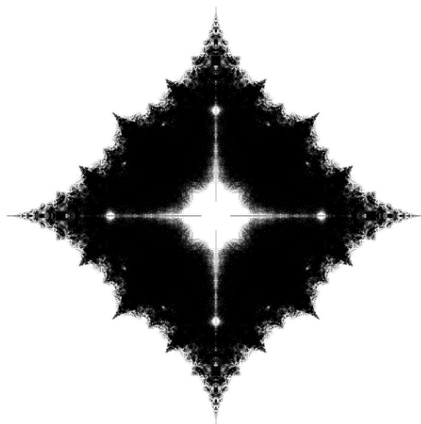
Size 19



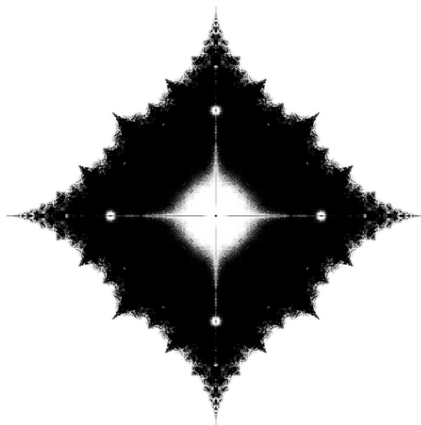
Size 20



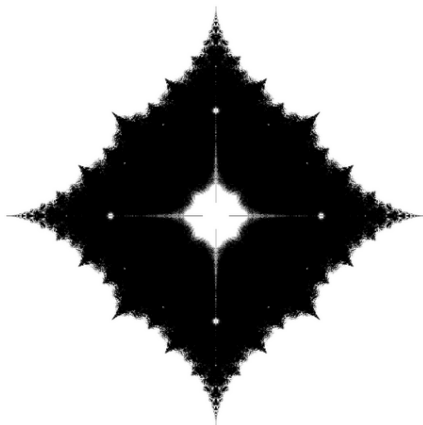
Size 21



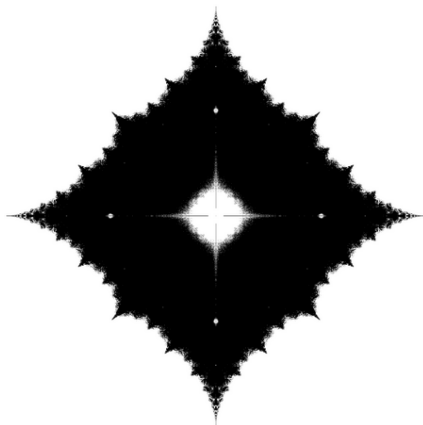
Size 22



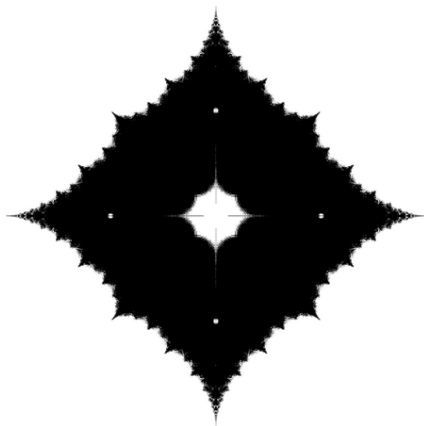
Size 23



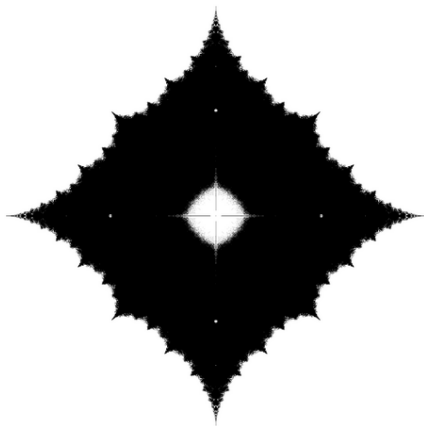
Size 24



Size 25



Size 26



Size 27

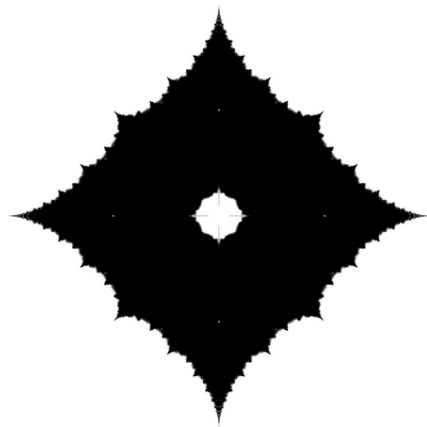


Size 28



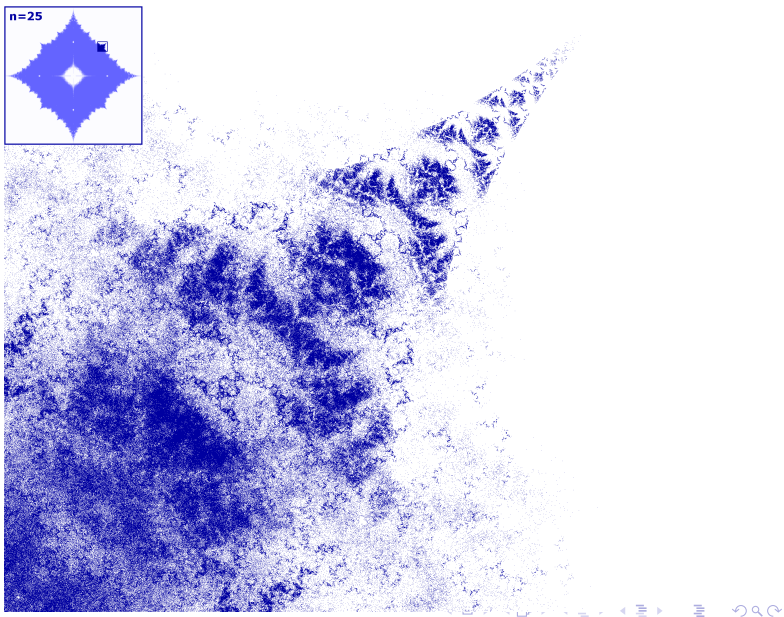
Size 29

Final Example [Feinberg/Zee 1999]



Size 30

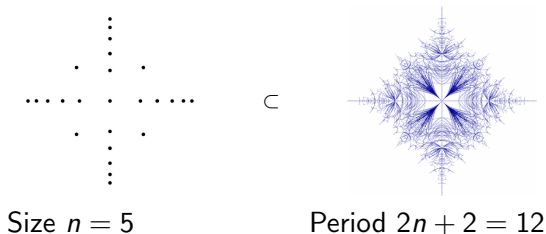
Zoom into Region $1 + i$ of σ_{25}^0



Final Example [Feinberg/Zee 1999]

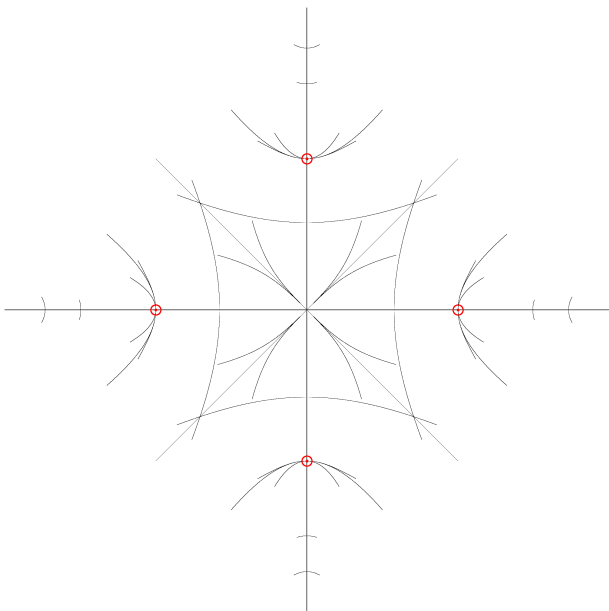
The **finite** matrix spectra σ_n^0 are even **contained** in the **periodic** (infinite) matrix spectra shown before.

More precisely, the spectra of all $n \times n$ principal submatrices are **contained** in the set of all $(2n+2)$ -periodic matrices:

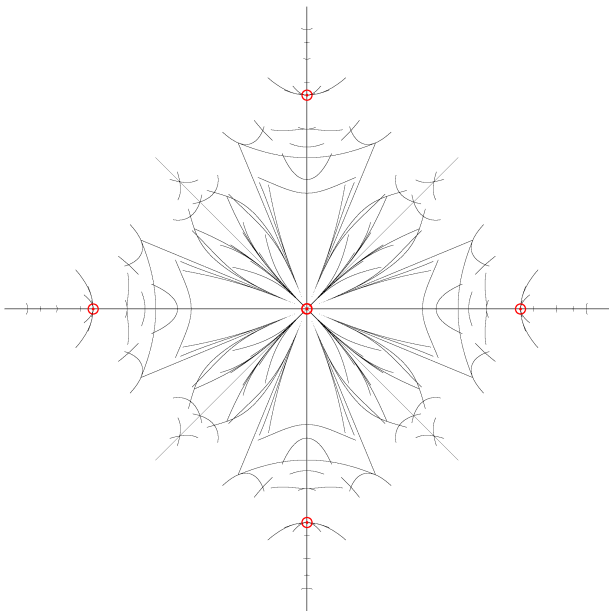


Here we demonstrate this inclusion for some values of n .

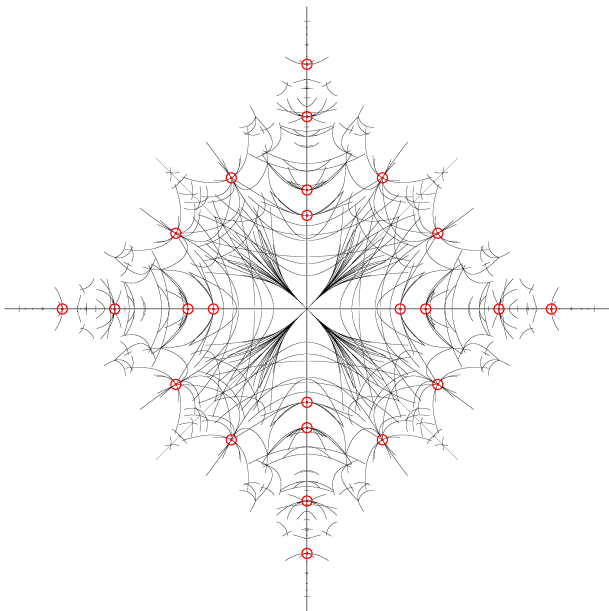
Finite Matrix Spectra in Periodic Matrix Spectra: $n = 2$



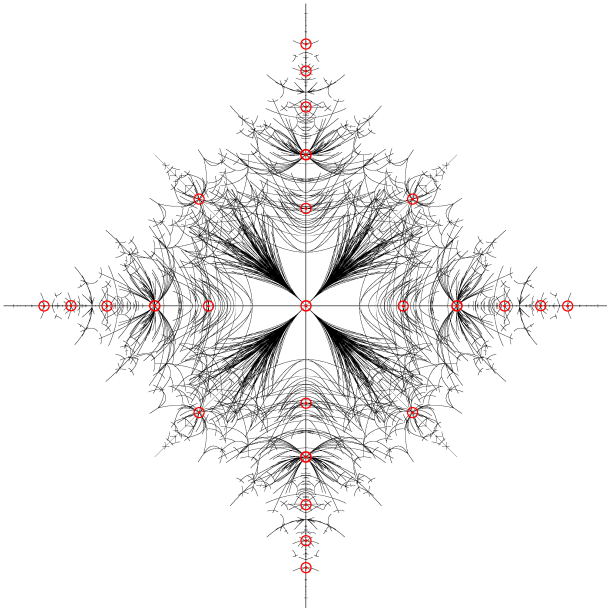
Finite Matrix Spectra in Periodic Matrix Spectra: $n = 3$



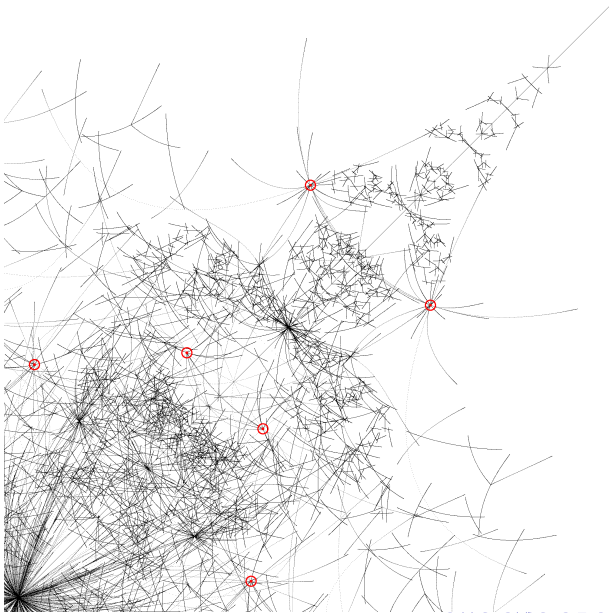
Finite Matrix Spectra in Periodic Matrix Spectra: $n = 4$



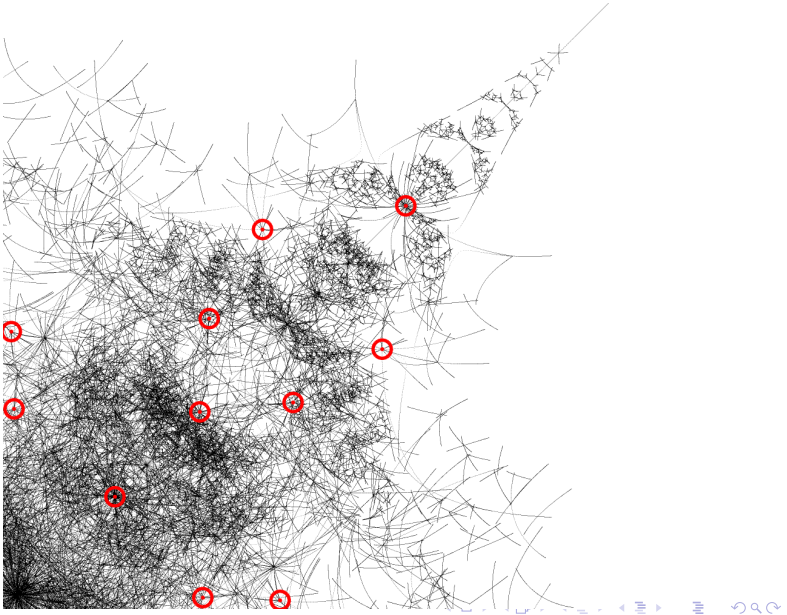
Finite Matrix Spectra in Periodic Matrix Spectra: $n = 5$



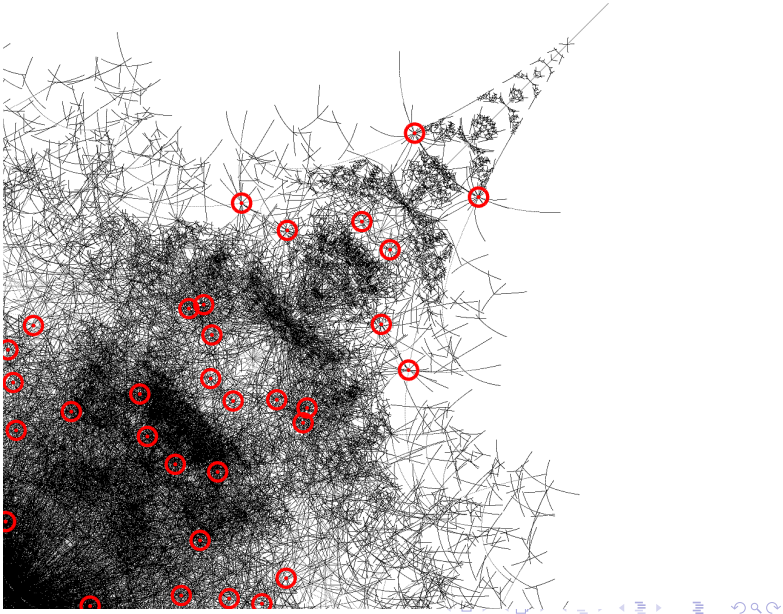
Finite Matrix Spectra in Periodic Matrix Spectra: $n = 8$



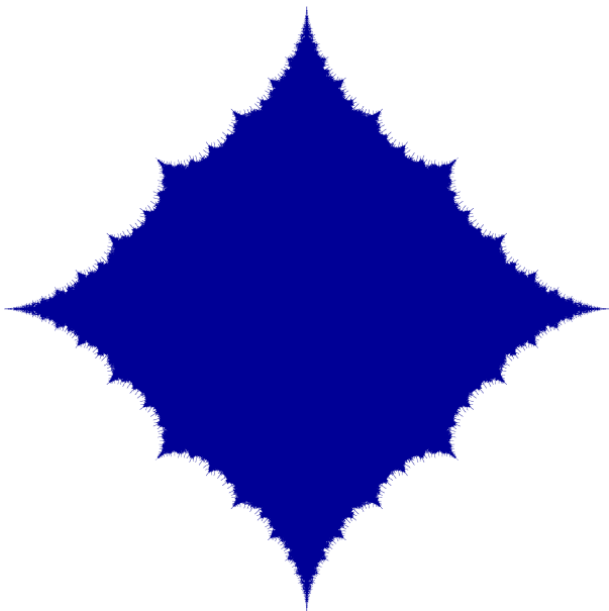
Finite Matrix Spectra in Periodic Matrix Spectra: $n = 9$



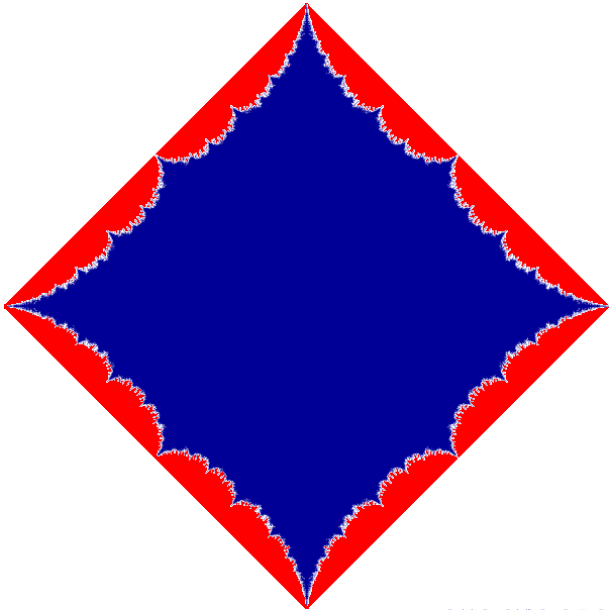
Finite Matrix Spectra in Periodic Matrix Spectra: $n = 10$



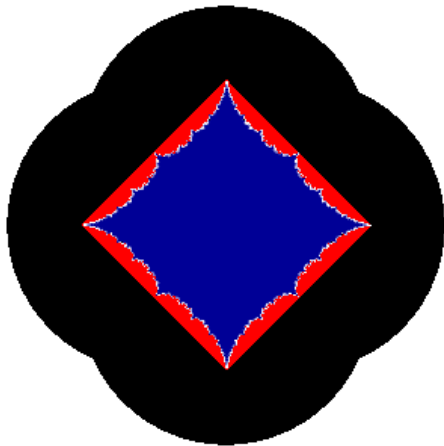
Conjecture: $\text{spec } A^b$ if b is pseudoergodic



Upper bound on $\text{spec } A^b$ by the closed numerical range

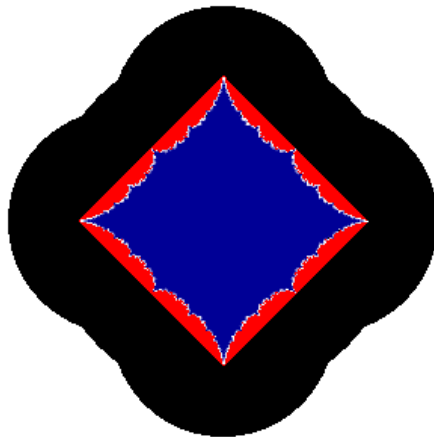


...in comparison with plots of $\Sigma_n^0 = \sigma_n^{0+\varepsilon_n}$



$n = 2$

...in comparison with plots of $\Sigma_n^0 = \sigma_n^{0+\varepsilon_n}$



$n = 3$

...in comparison with plots of $\Sigma_n^0 = \sigma_n^{0+\varepsilon_n}$



$n = 4$

...in comparison with plots of $\Sigma_n^0 = \sigma_n^{0+\varepsilon_n}$



$n = 5$

...in comparison with plots of $\Sigma_n^0 = \sigma_n^{0+\varepsilon_n}$



$n = 6$

...in comparison with plots of $\Sigma_n^0 = \sigma_n^{0+\varepsilon_n}$



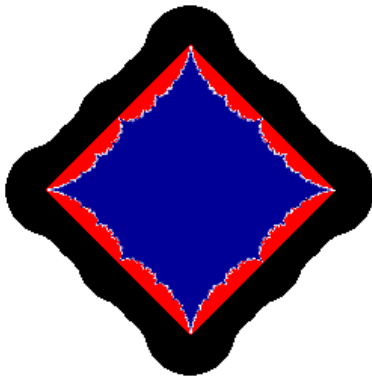
$n = 7$

...in comparison with plots of $\Sigma_n^0 = \sigma_n^{0+\varepsilon_n}$



$n = 8$

...in comparison with plots of $\Sigma_n^0 = \sigma_n^{0+\varepsilon_n}$



$n = 9$

...in comparison with plots of $\Sigma_n^0 = \sigma_n^{0+\varepsilon_n}$



$n = 10$

...in comparison with plots of $\Sigma_n^0 = \sigma_n^{0+\varepsilon_n}$



$n = 11$

...in comparison with plots of $\Sigma_n^0 = \sigma_n^{0+\varepsilon_n}$



$$n = 12$$

...in comparison with plots of $\Sigma_n^0 = \sigma_n^{0+\varepsilon_n}$



$n = 13$

...in comparison with plots of $\Sigma_n^0 = \sigma_n^{0+\varepsilon_n}$



$n = 14$

...in comparison with plots of $\Sigma_n^0 = \sigma_n^{0+\varepsilon_n}$



$n = 15$

...in comparison with plots of $\Sigma_n^0 = \sigma_n^{0+\varepsilon_n}$



$n = 16$

...in comparison with plots of $\Sigma_n^0 = \sigma_n^{0+\varepsilon_n}$



$n = 17$

...in comparison with plots of $\Sigma_n^0 = \sigma_n^{0+\varepsilon_n}$



$n = 18$

Where does \sum_n^0 go as $n \rightarrow \infty$?

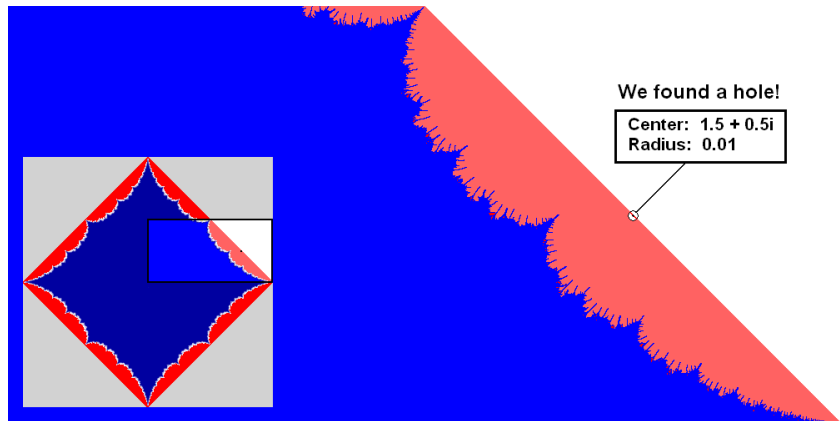
Computational cost for these pics: $n \cdot 2^{n-1} \times$ number of pixels.

Where does \sum_n^0 go as $n \rightarrow \infty$?

Computational cost for these pics: $n \cdot 2^{n-1} \times$ number of pixels.
So let us focus on just **one** point (pixel) λ :

Where does Σ_n^0 go as $n \rightarrow \infty$?

Computational cost for these pics: $n \cdot 2^{n-1} \times \text{number of pixels}$.
So let us focus on just **one** point (pixel) λ :



$$\lambda = 1.5 + 0.5i \notin \Sigma_{36}^0 \supset \text{spec } A^b$$

Beyond Jacobi Matrices

- 1 Introduction
- 2 Inclusion Sets and Approximations for the Spectrum and Pseudospectrum
- 3 Substantial example with lots of pictures! Feinberg-Zee Random Hopping Matrix
- 4 Extensions

- Everything goes through if \mathbb{C} replaced by $B(X)$
- So everything goes through for matrices with arbitrary bandwidth
- Small perturbations can be handled



S.N. CHANDLER-WILDE and M. LINDNER:
Limit Operators, Collective Compactness & Spectral Theory ...,
Volume 210 (Number 989) of Memoirs of the AMS, 2011.



S.N. CHANDLER-WILDE and M. LINDNER:
Sufficiency of Favard's Condition for a class of ...
Journal of Functional Analysis 2008



C-W, R. CHONCHAIYA and M. LINDNER:
Eigenvalue problem meets Sierpinski triangle ...
Operators and Matrices 2011



S.N. CHANDLER-WILDE and E.B. DAVIES:
Spectrum of a Feinberg-Zee random hopping matrix
Journal of Spectral Theory 2012



C-W, R. CHONCHAIYA and LINDNER:
On the spectra and pseudospectra of a class of NSA random ...
Operators and Matrices 2013